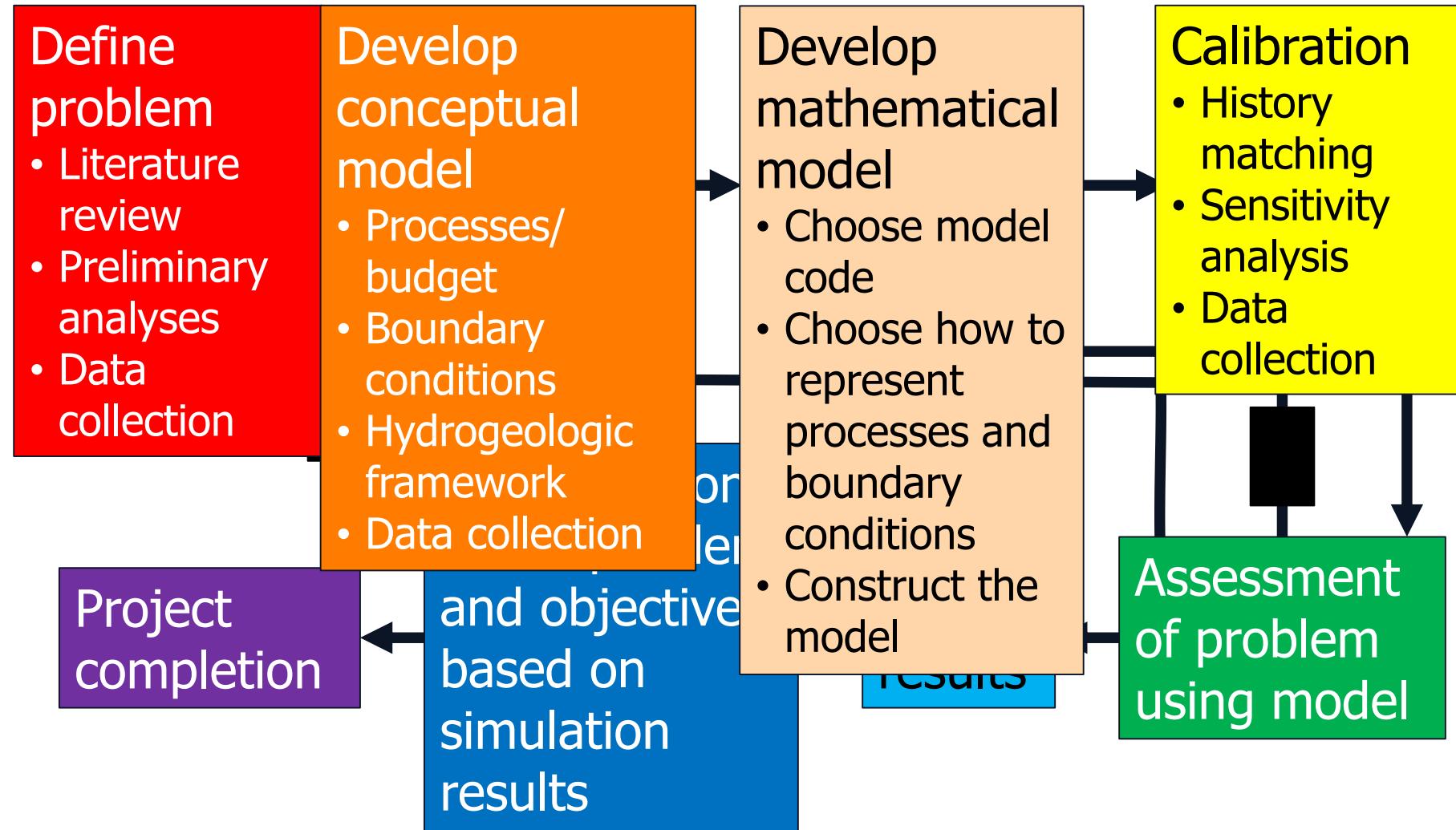


# Model Calibration

Stephen Hundt

# The modeling process

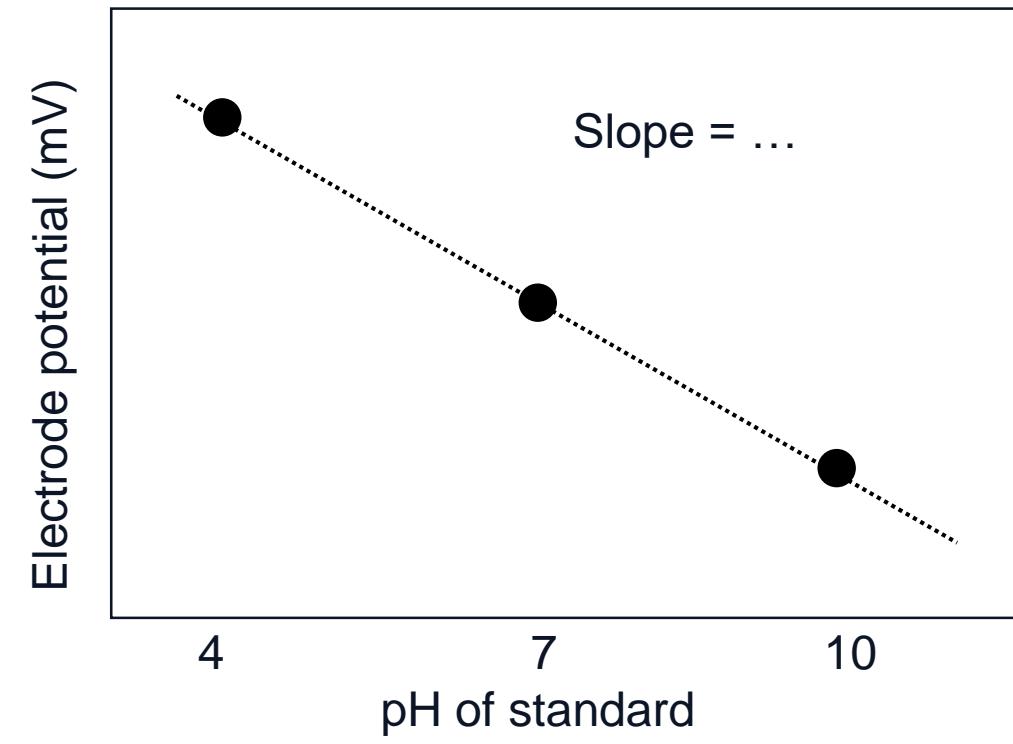


After Reilly (2001) TWRI 3,B8

# What is Model Calibration?

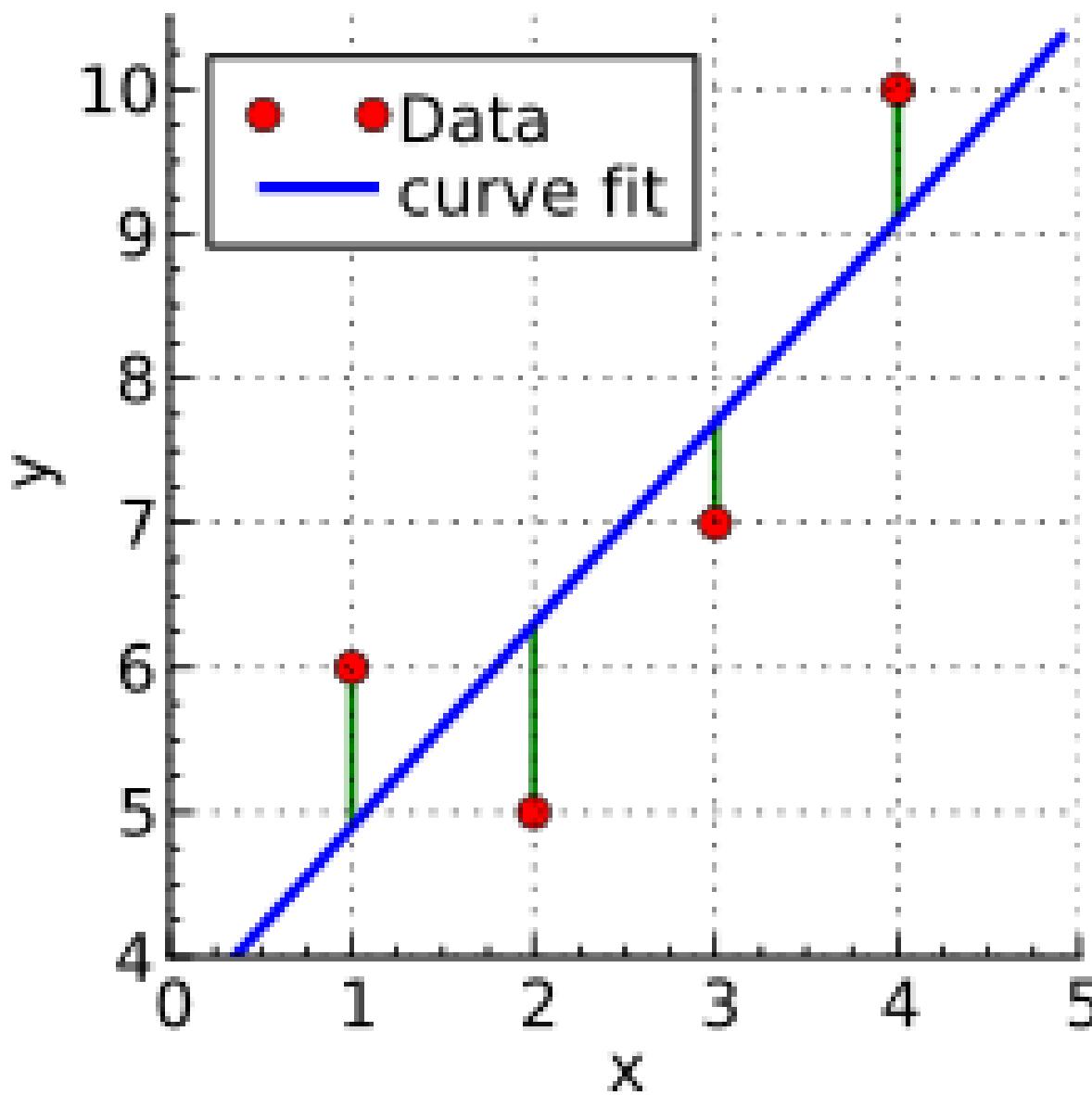
# Other Disciplines

## instrumentation 'calibration'



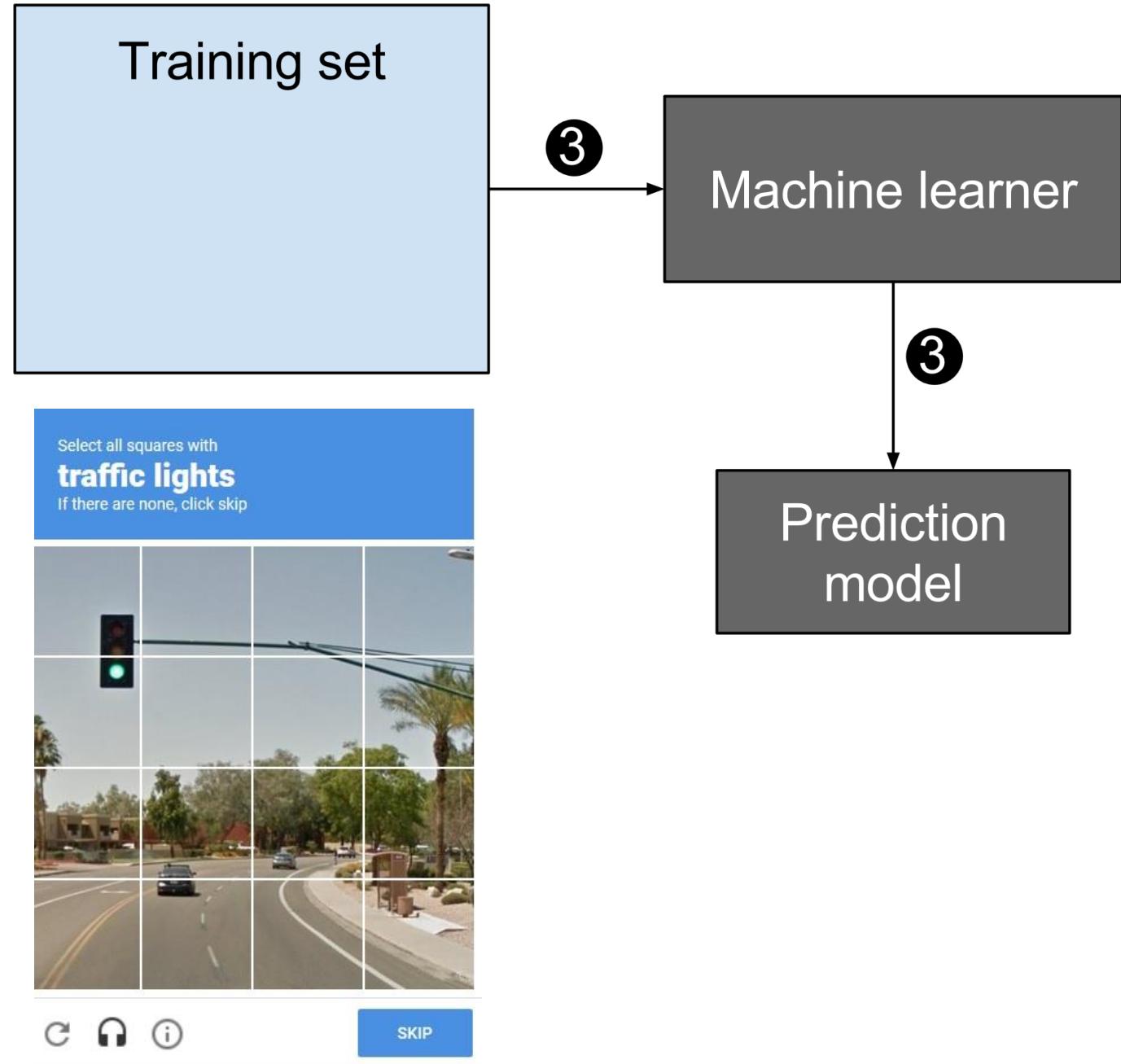
# Other Disciplines

'fitting'  
curves



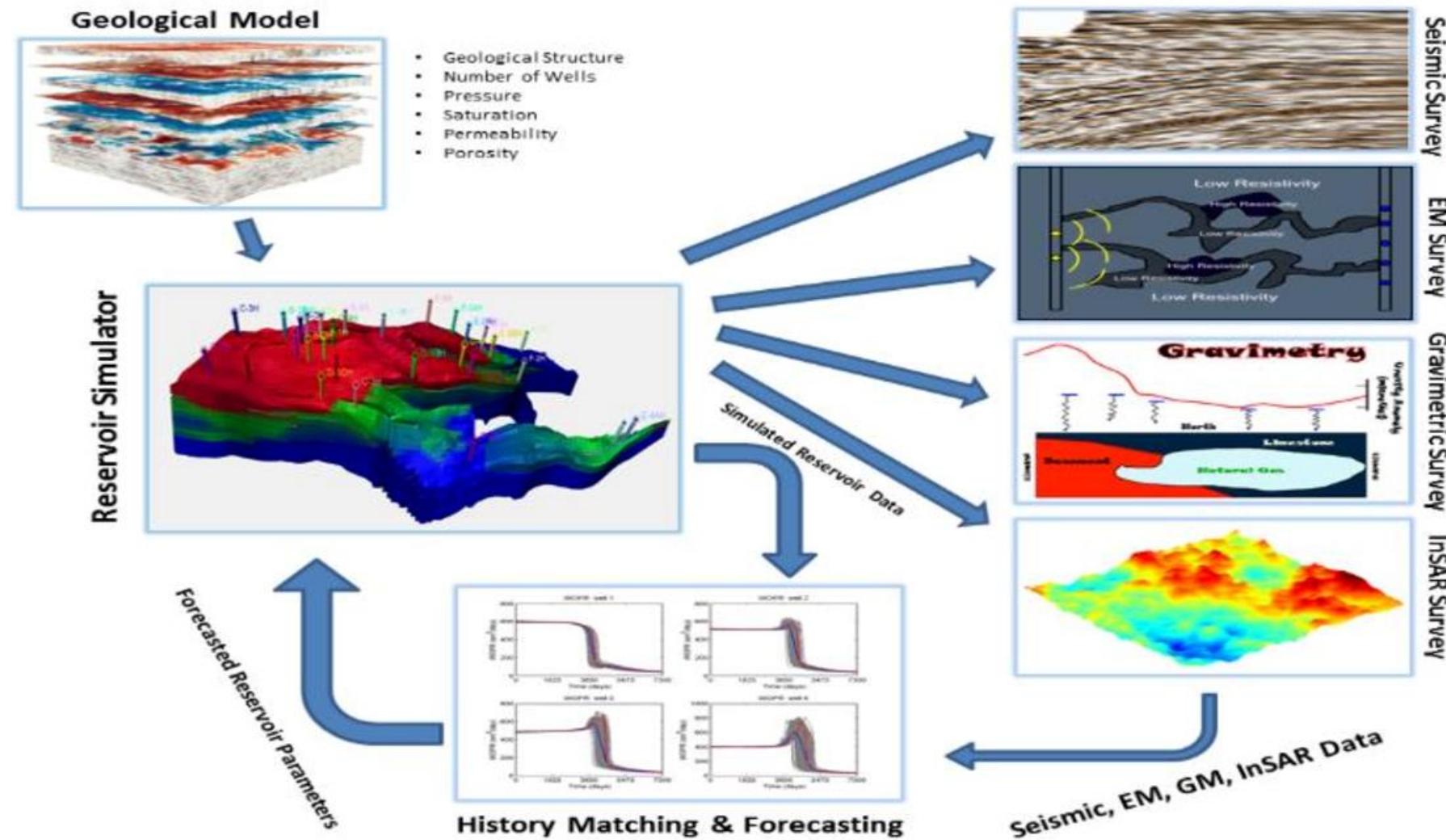
# Other Disciplines

'training' machine learning models



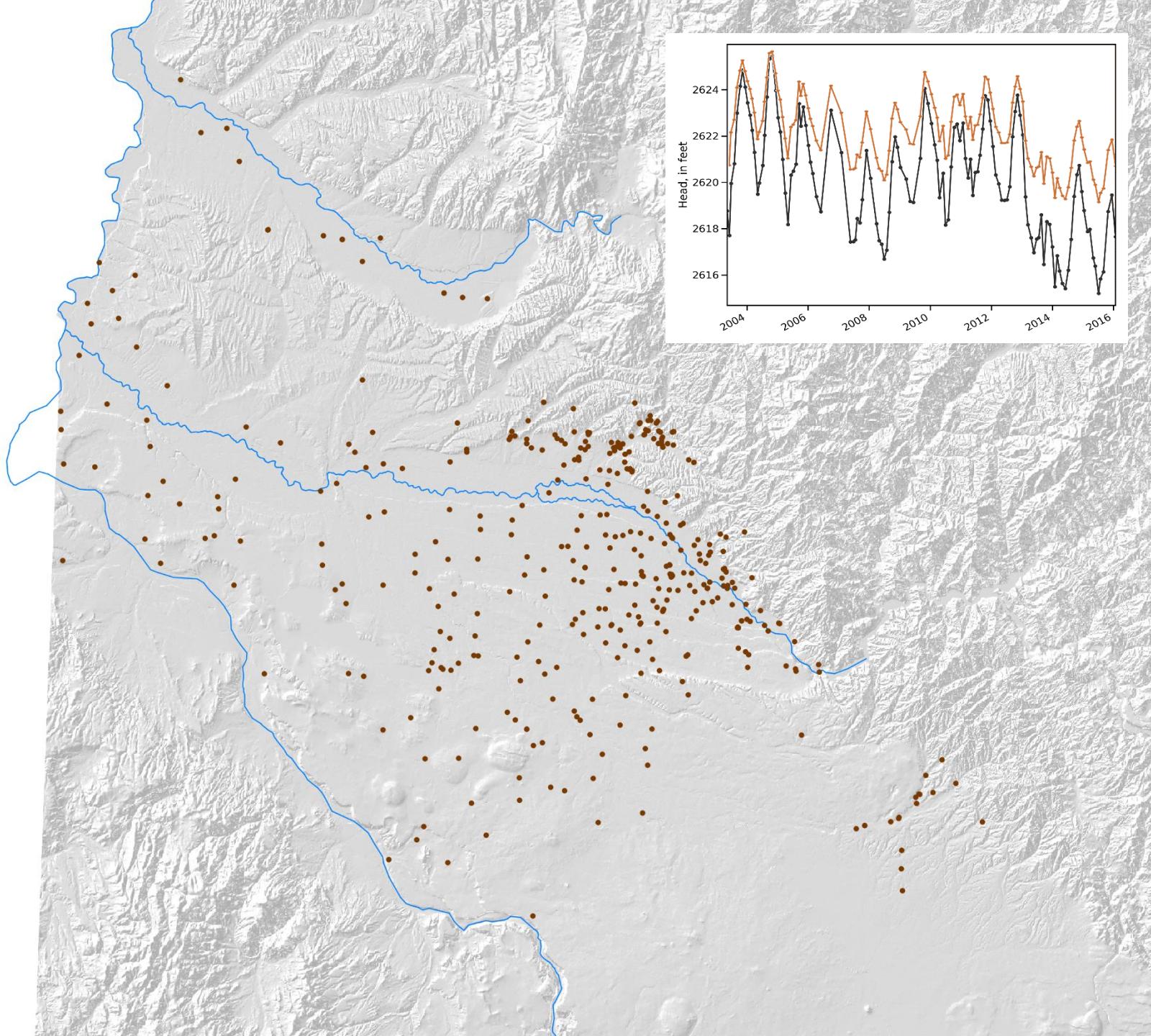
# Other Disciplines

'history-matching' petroleum reservoir models



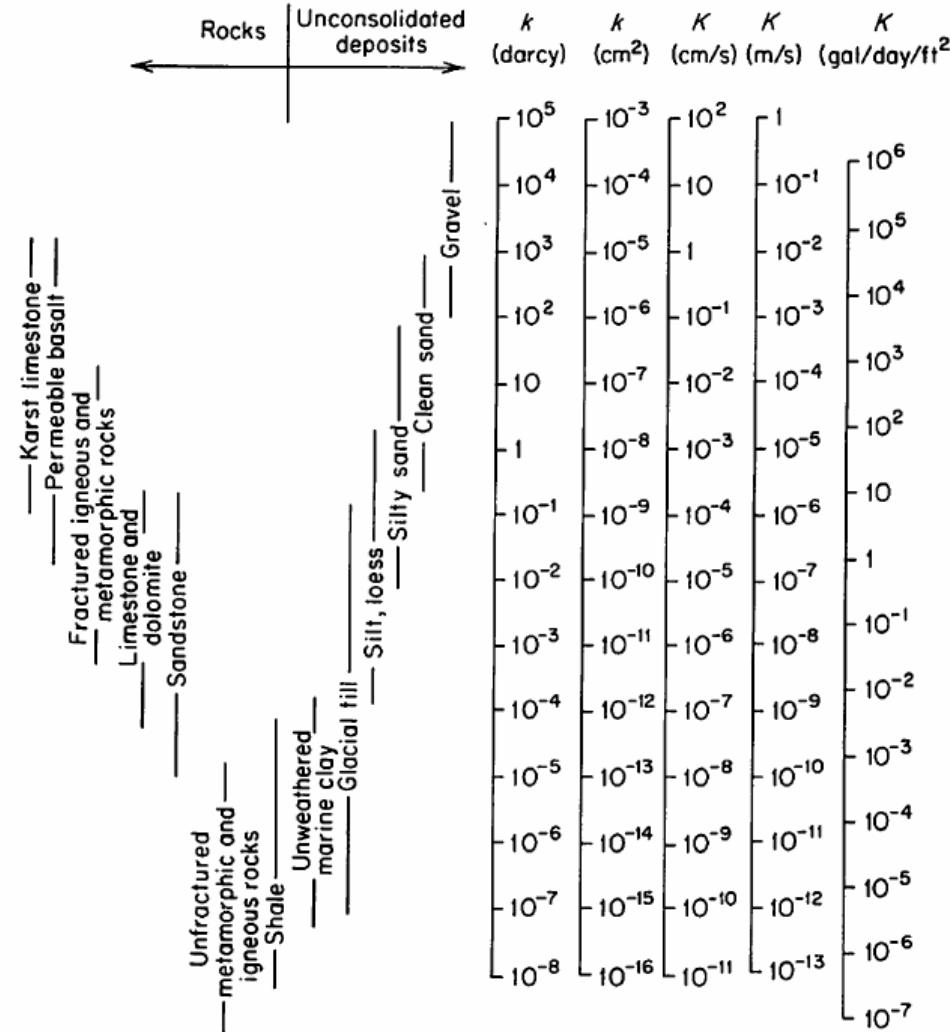
# Groundwater Modelling

Calibration:  
History-matching,  
Parameter  
estimation,  
Inverse modelling



# Why do we calibrate?

# Unknown Material Properties

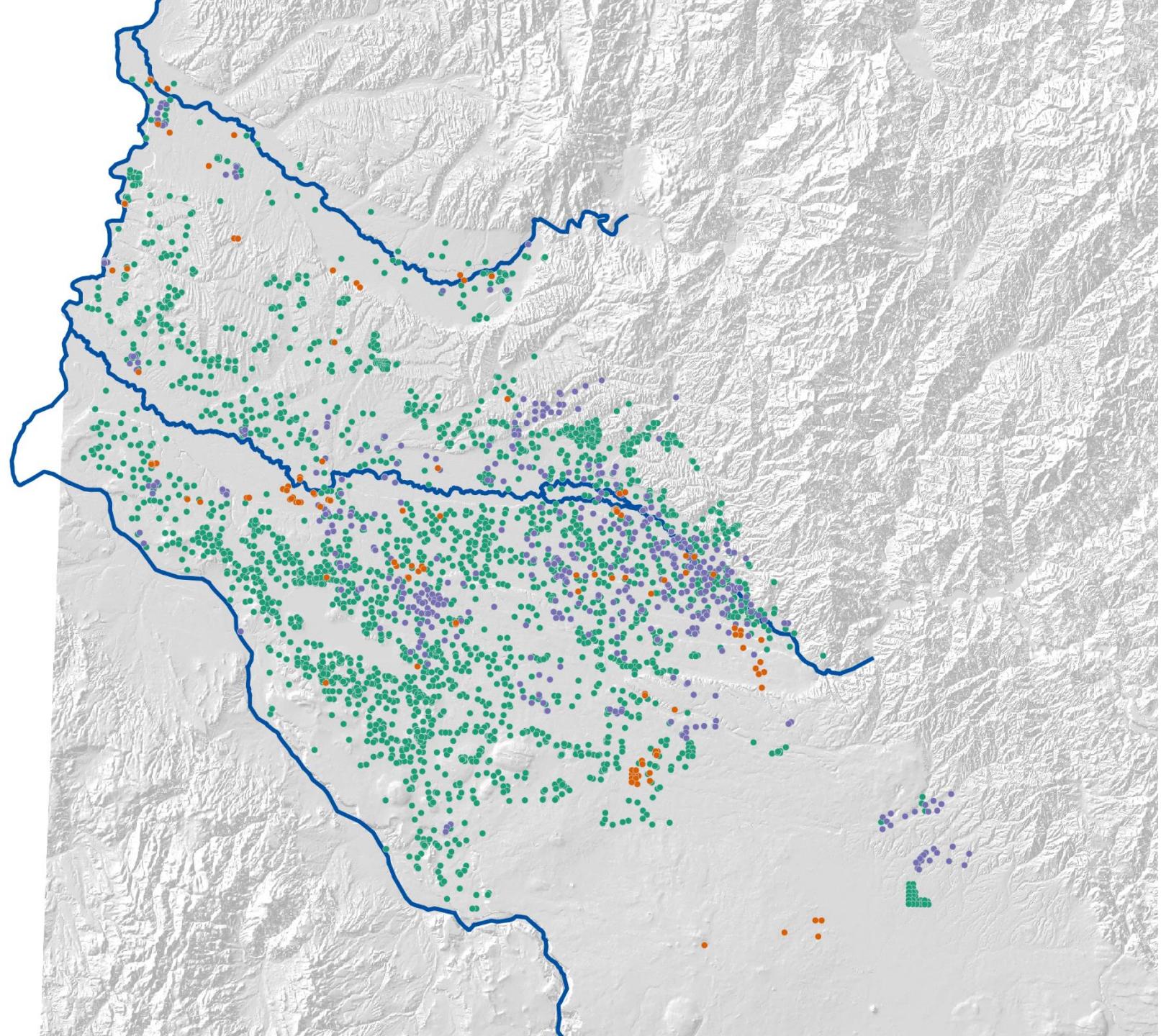


**13. LITHOLOGIC LOG: (Describe repairs or abandonment)**

Bore Dia.	From	To	Remarks: Lithology, Water Quality & Temperature	Y	N
18"	0'	2'	brown top soil		X
18"	2'	4'	brown clay		X
18"	4'	27'	brown gravel		X
18"	27'	46'	brown clay		X
18"	46'	54'	brown sand & gravel		X
18"	54'	61'	brown clay		X
18"	61'	89'	brown sand & gravel		X
18"	89'	91'	brown sandy clay		X
18"	91'	98'	brown sand & gravel		X
18"	98'	108	brown sand fine		X
18"	108	114	brown sand & clay strips		X
18"	114	127	brown sand fine		X
18"	127	130	brown clay		X
18"	130	143	brown sand & clay strips		X

**RECEIVED**

# Unknown Boundary Fluxes

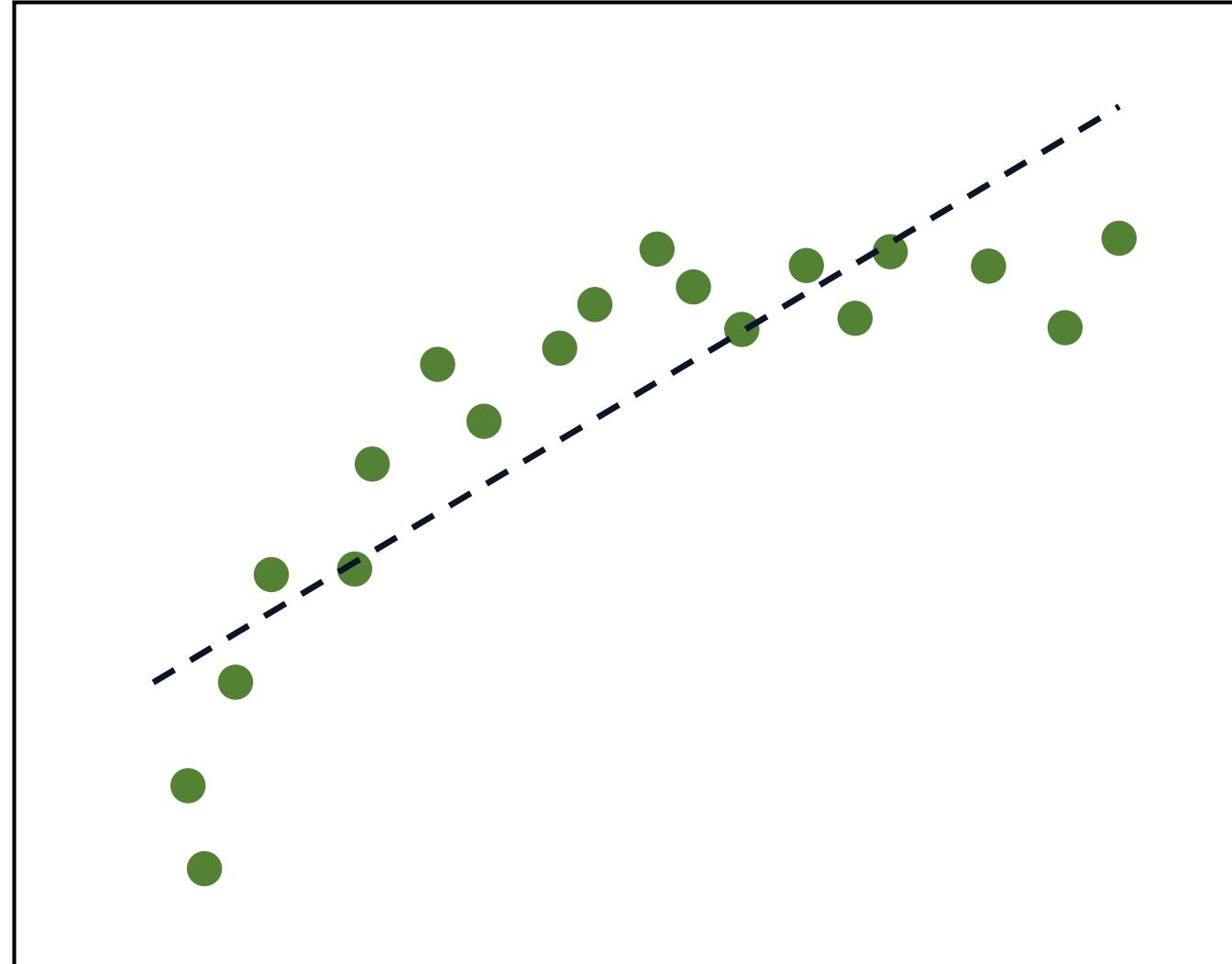


# Model Structural Error

Model represents a  
process ‘wrong’

Too simple?

Parameter  
compensation of  
model error



# Unavoidable?

Simple

Complex



too much  
structural  
error

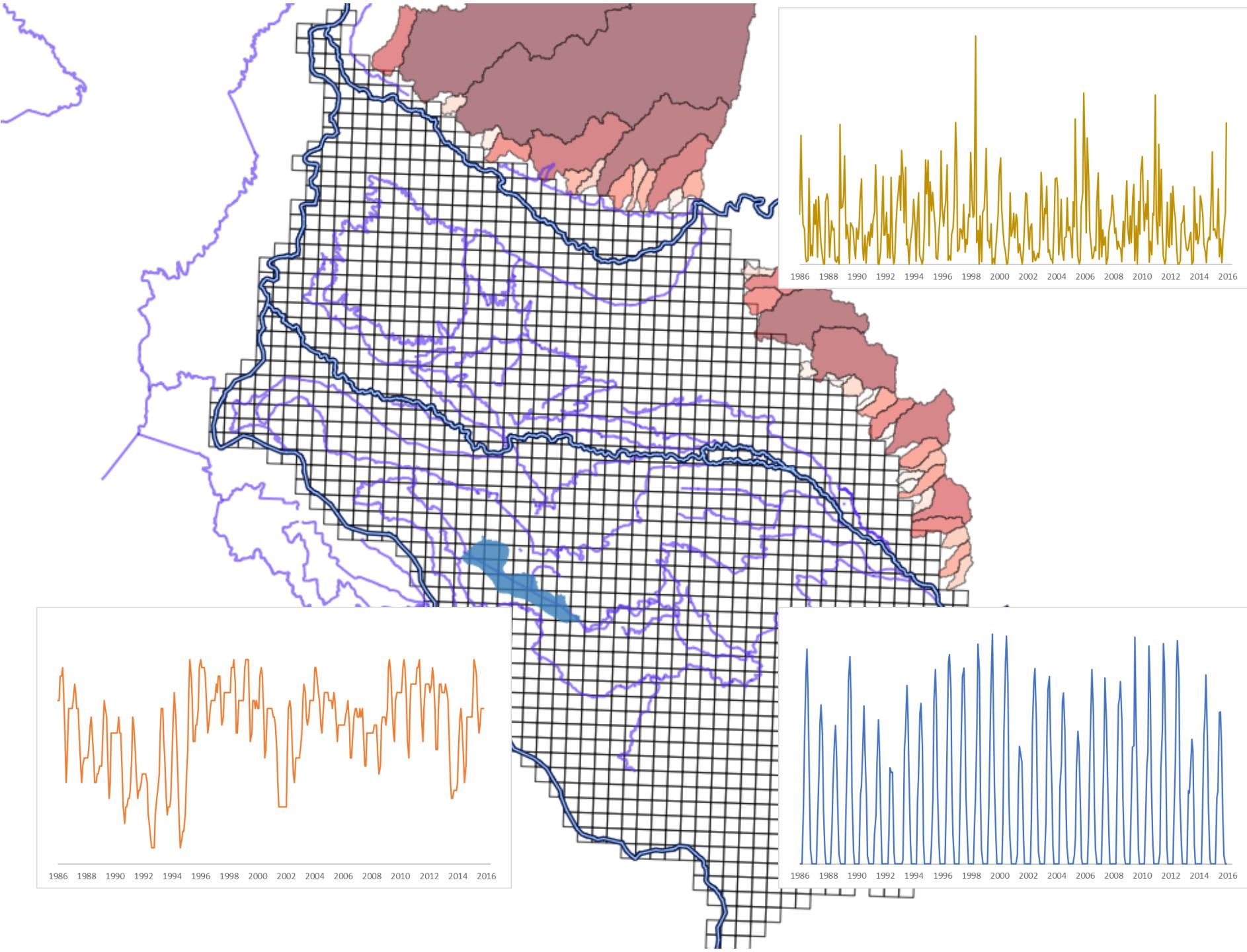
too time  
consuming  
and unstable  
to calibrate



- more resolution
- more parameters
- more processes
- more realistic process representation
- less numerically stable
- longer run times
- ...still structural error

# How do we calibrate?

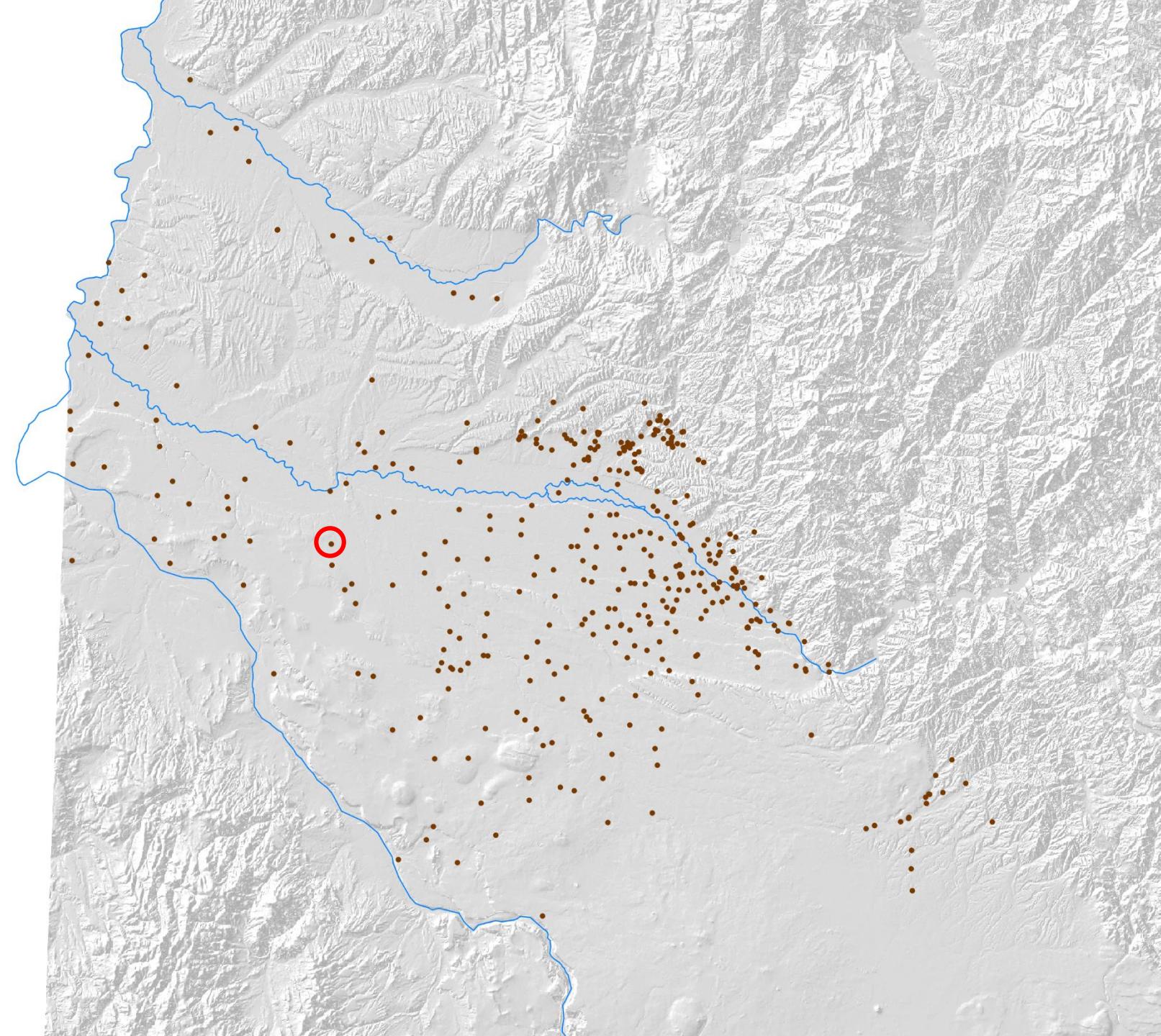
# Construct Historic Scenario



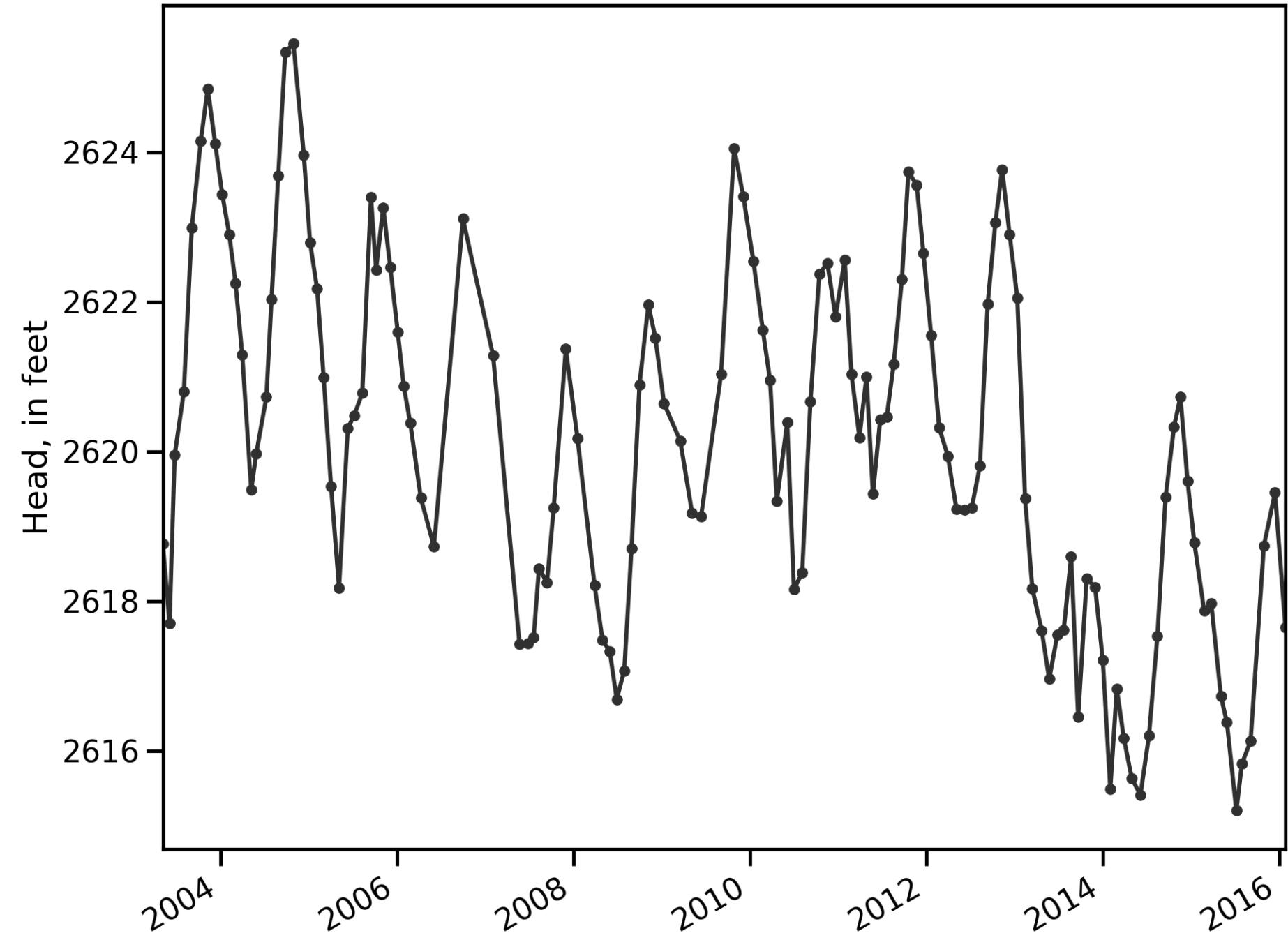
# Field Observations



# Field Observations



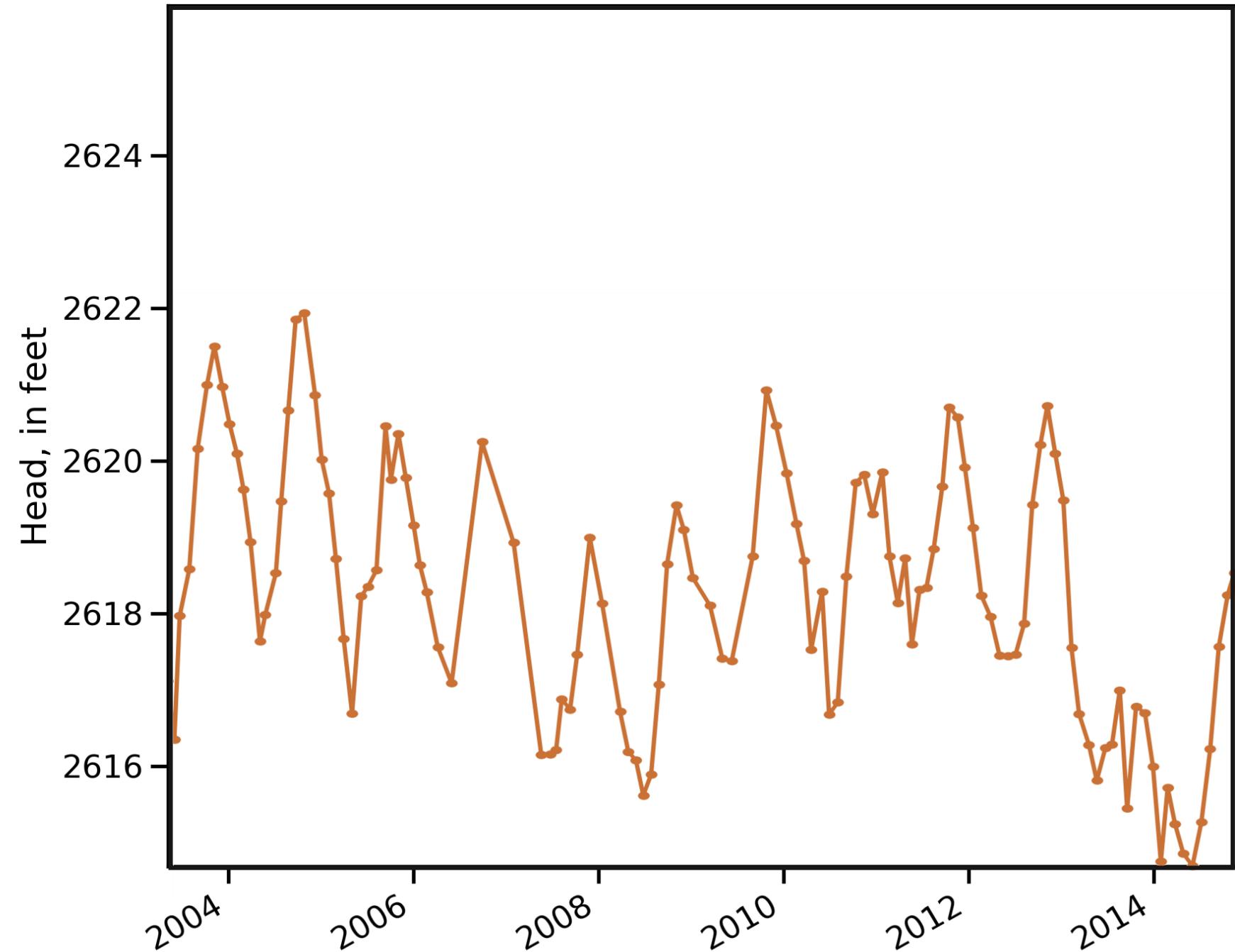
# Field Observations



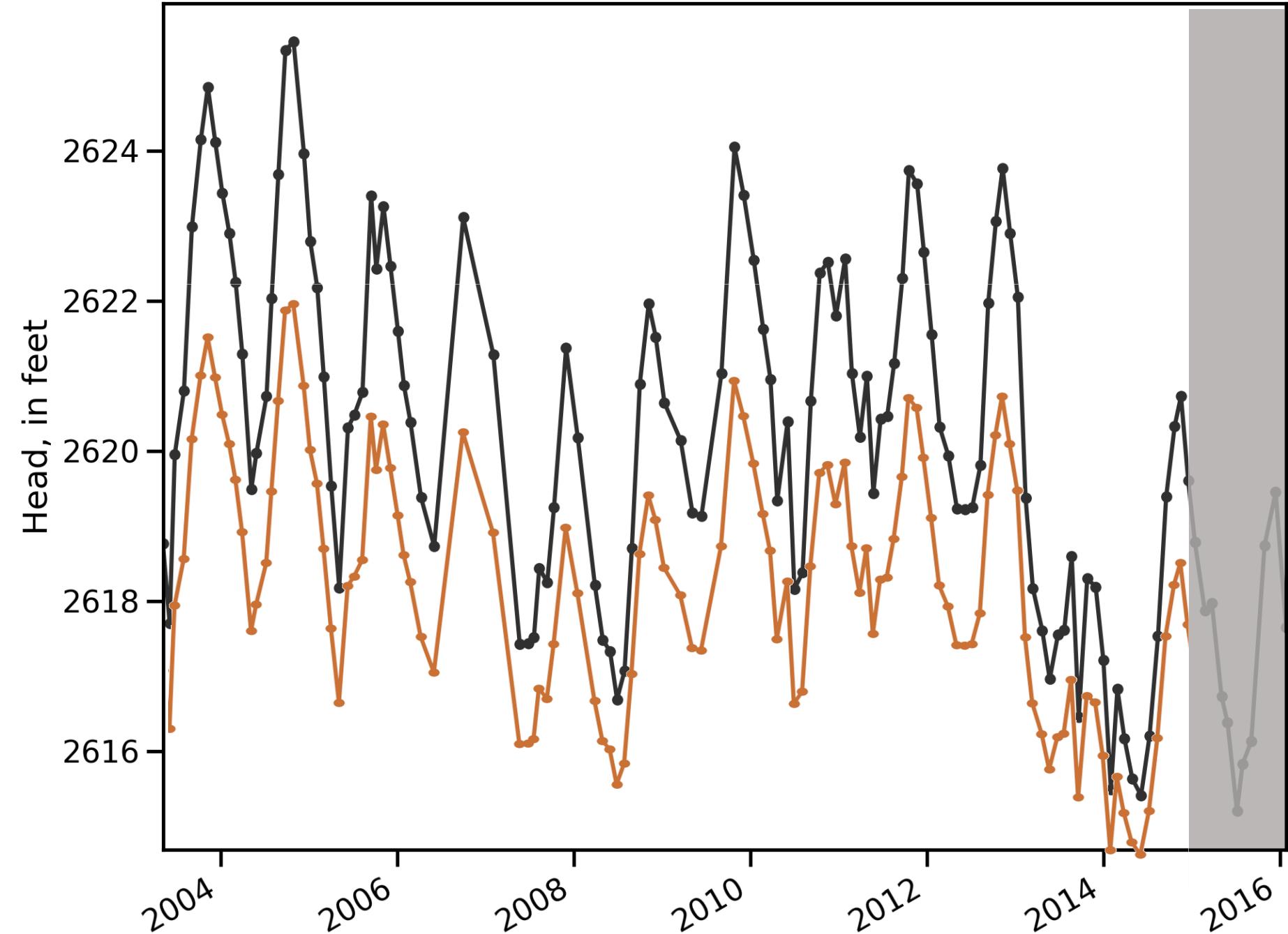
# Simulated Equivalents



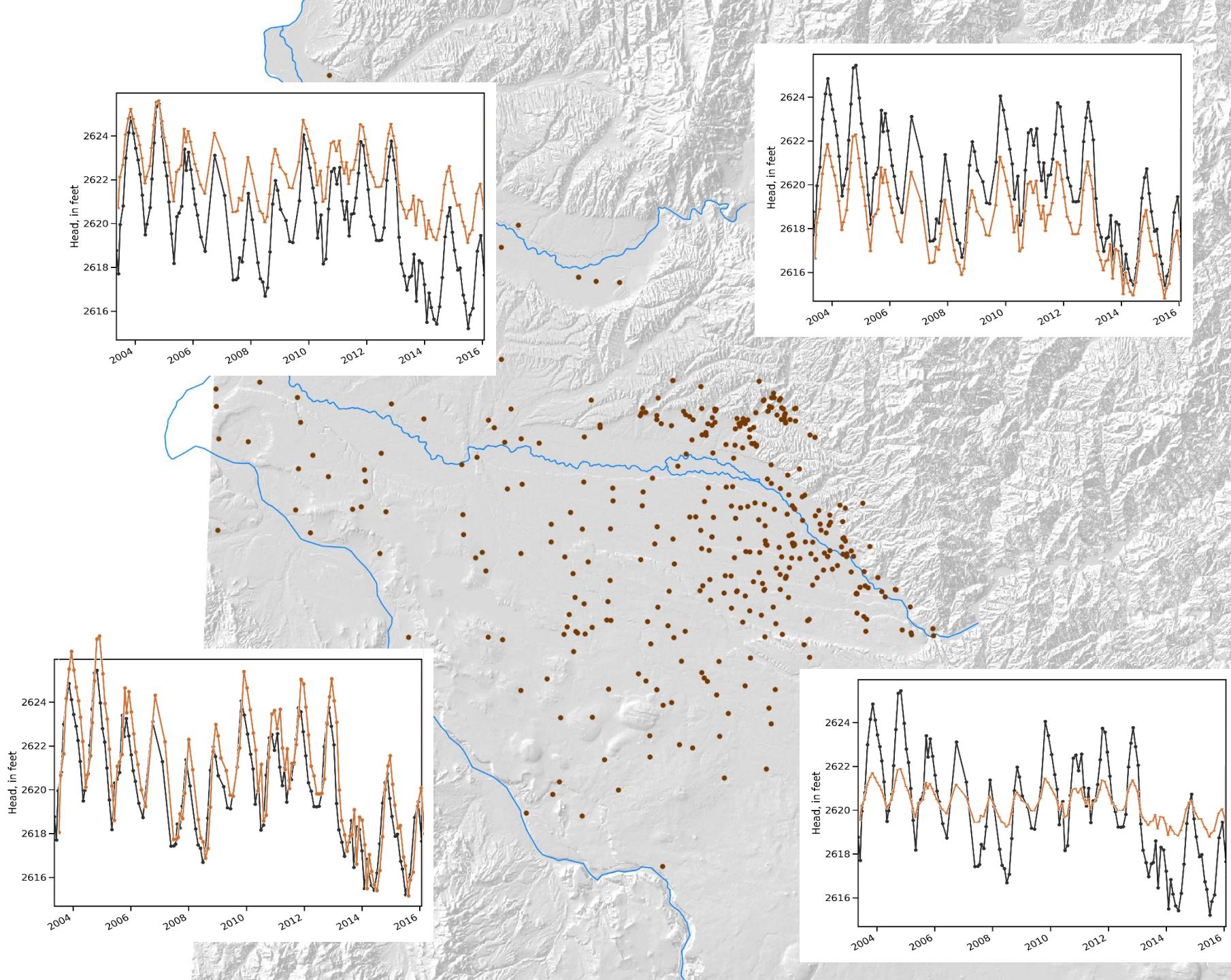
# Simulated Equivalents



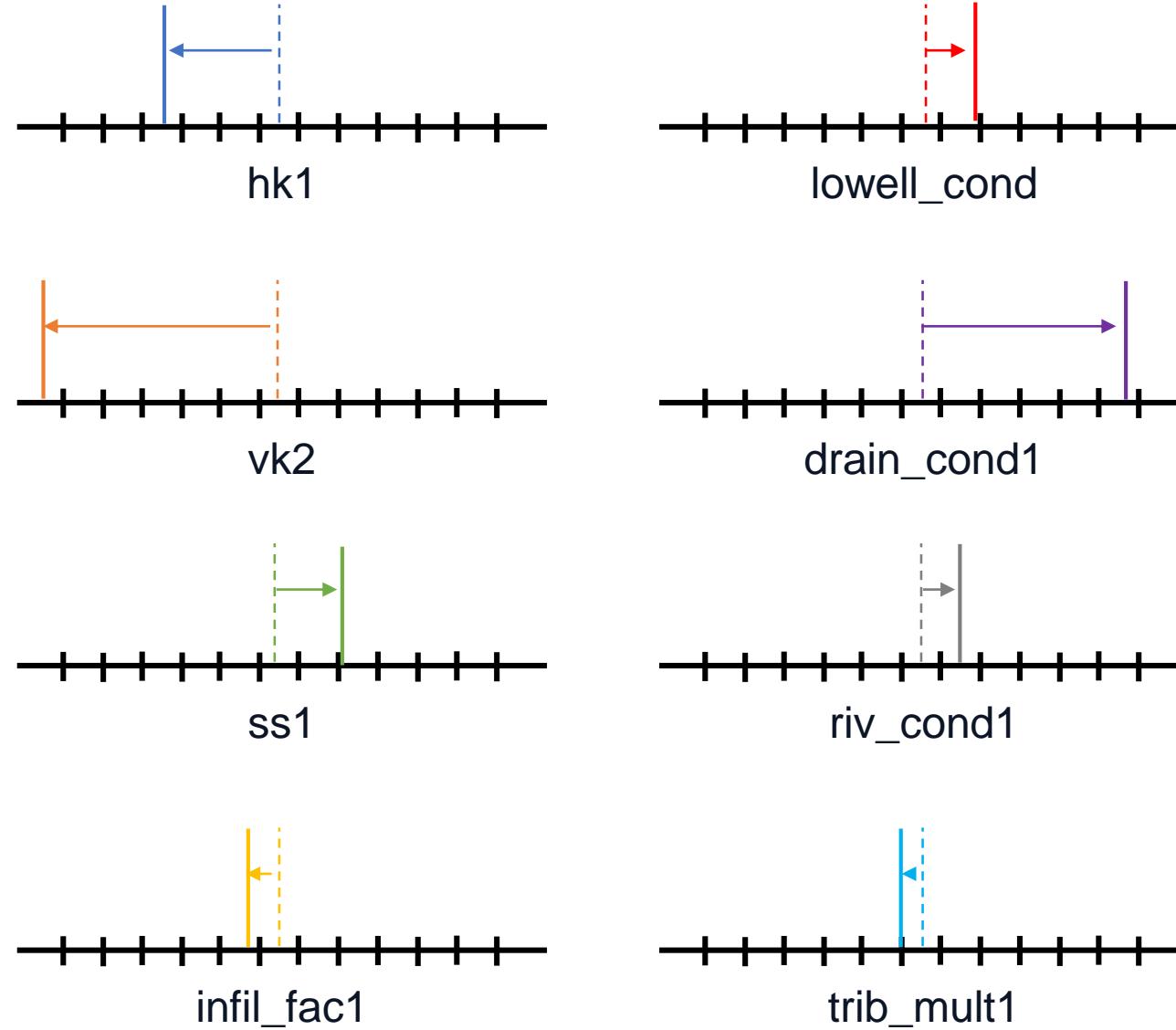
# Observed compared to simulated



# Observed compared to simulated

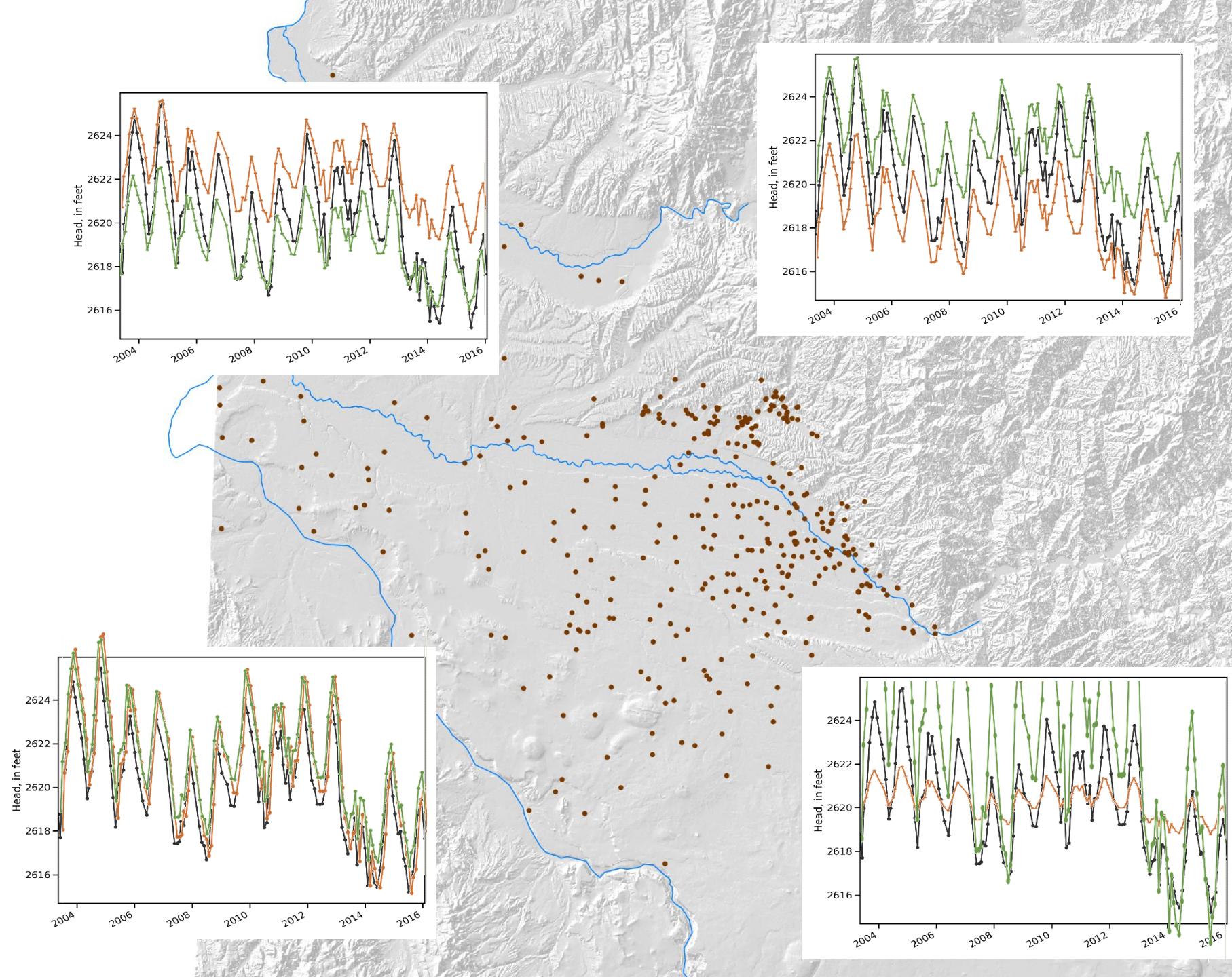


# Tweak Parameter Values

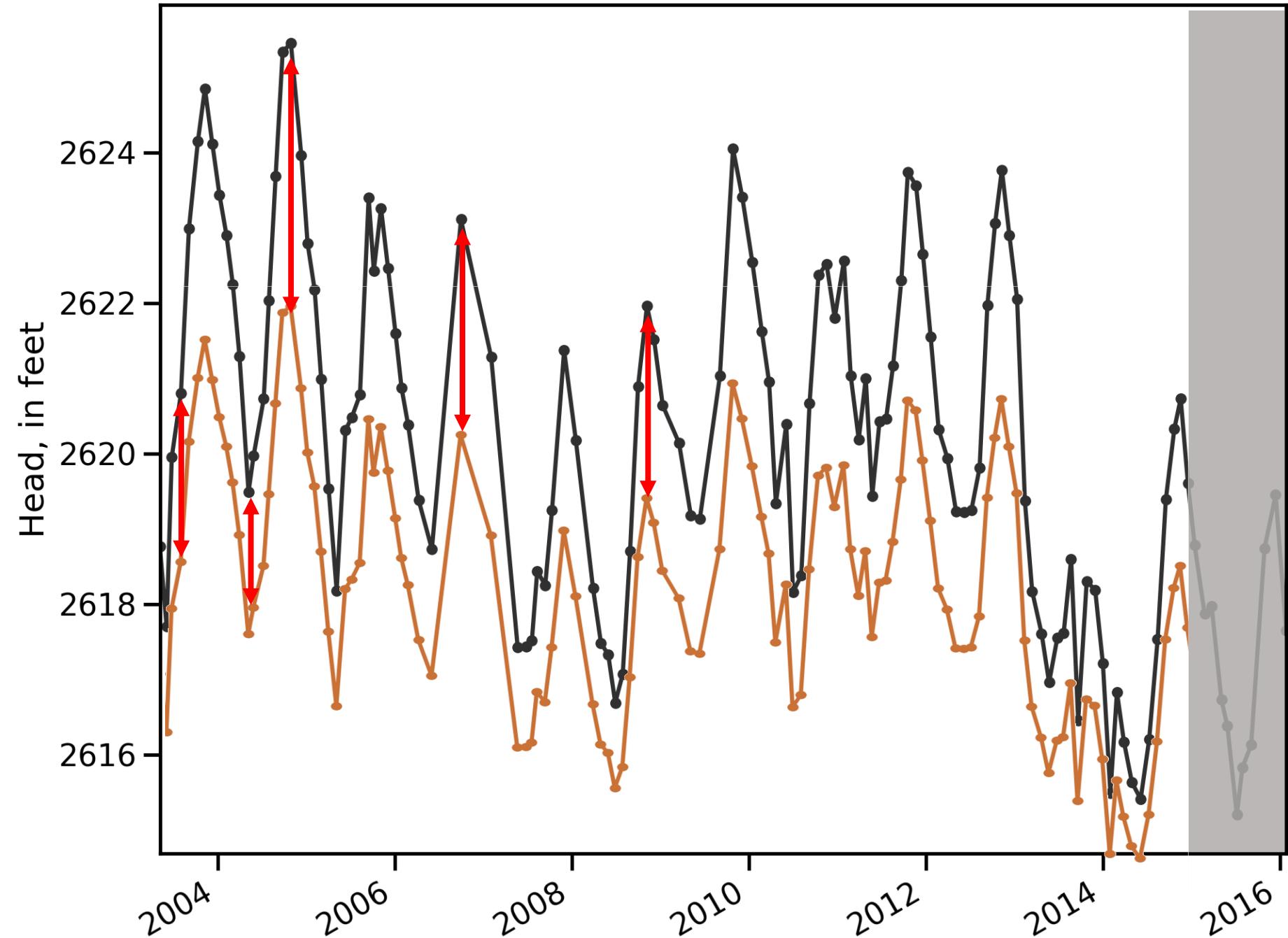


and so on...

# Tweak Parameter Values

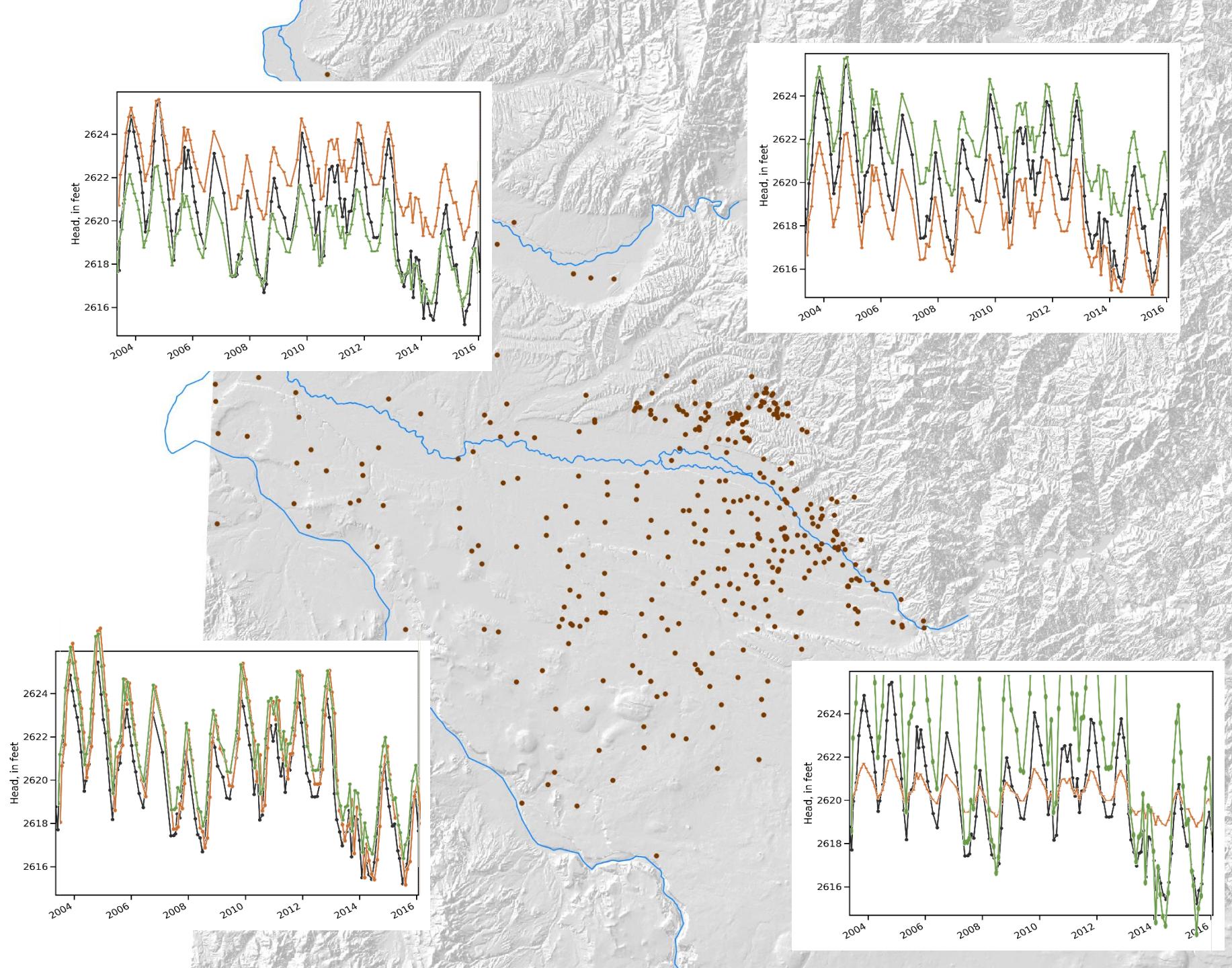


# Residuals



# Residuals

Residuals = 2, -3, 4, ...



# Objective Function

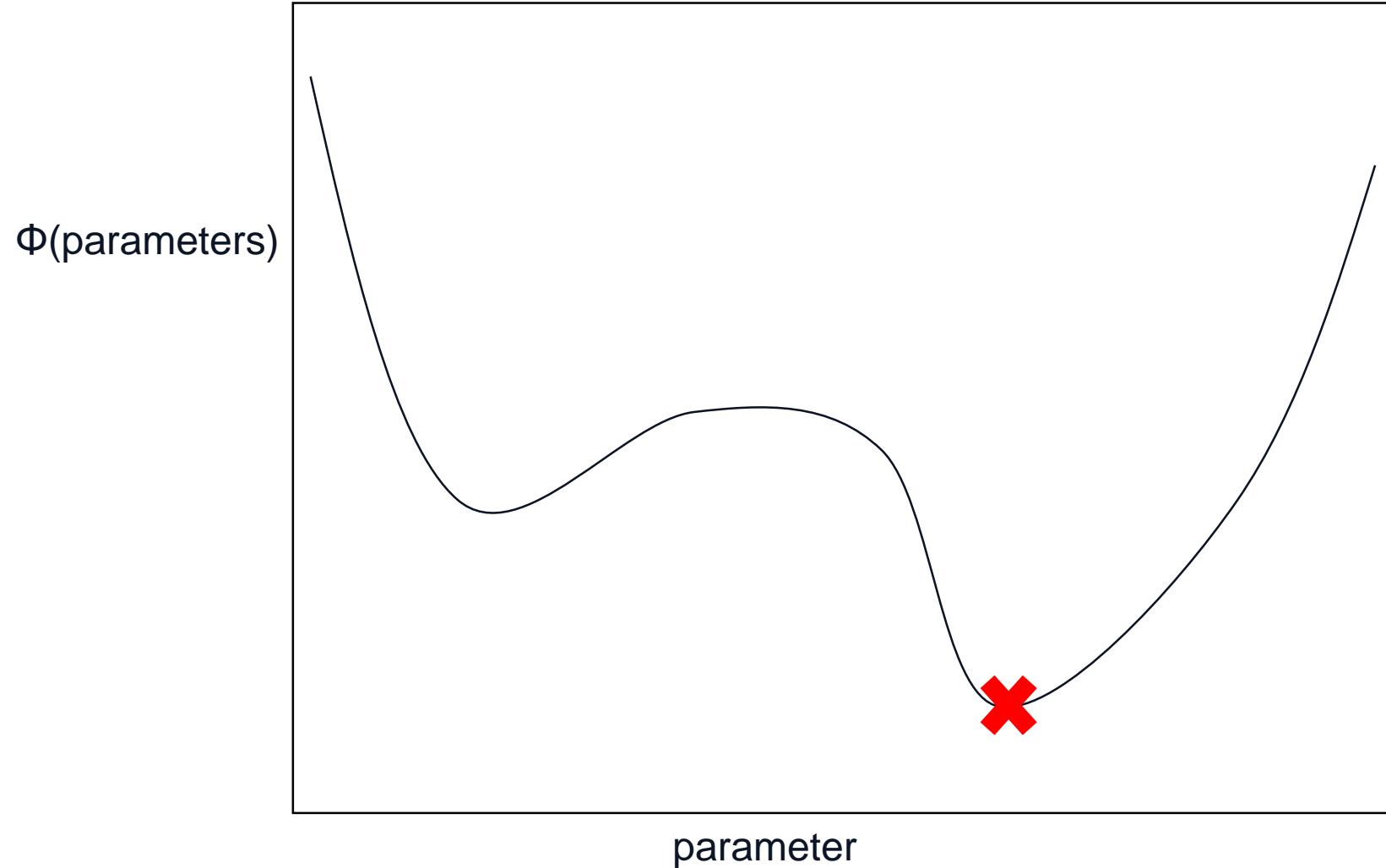
$$r_i = \text{observed}_i - \text{simulated}_i$$

$$w_i = \text{weight}$$

$$\text{objective function: } \Phi = \sum (w_i r_i)^2$$

$$\begin{aligned}\Phi &= \sum (w_{heads} r_{heads})^2 \\ &+ \sum (w_{vert\_head\_diff} r_{vert\_head\_diff})^2 \\ &+ \sum (w_{temp\_head\_diff} r_{temp\_head\_diff})^2 \\ &+ \sum (w_{drain\_flux} r_{drain\_flux})^2 \\ &+ \sum (w_{lowell\_seepage} r_{lowell\_seepage})^2\end{aligned}$$

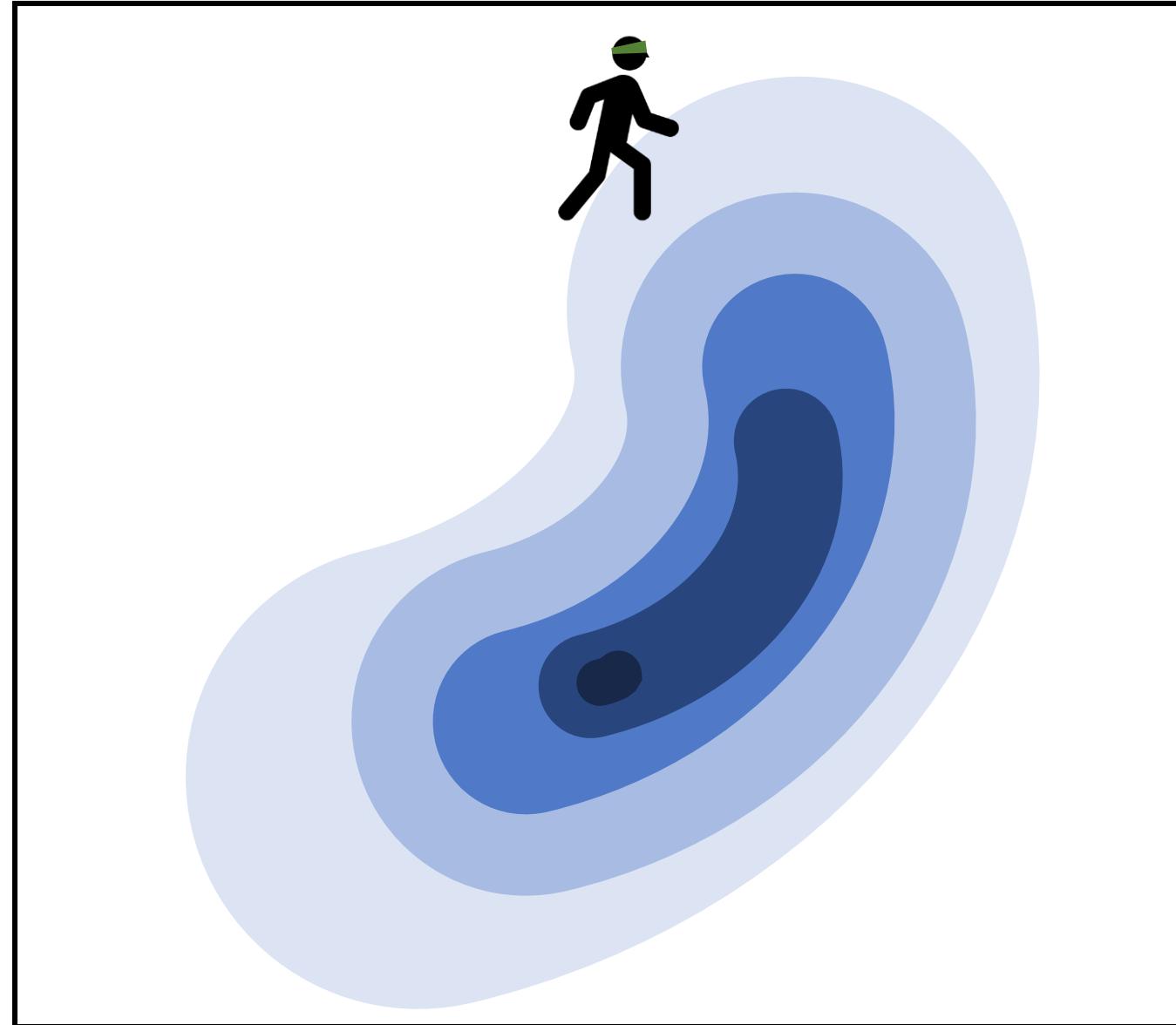
# Minimizing Objective Function



# Minimizing Objective Function

North - South

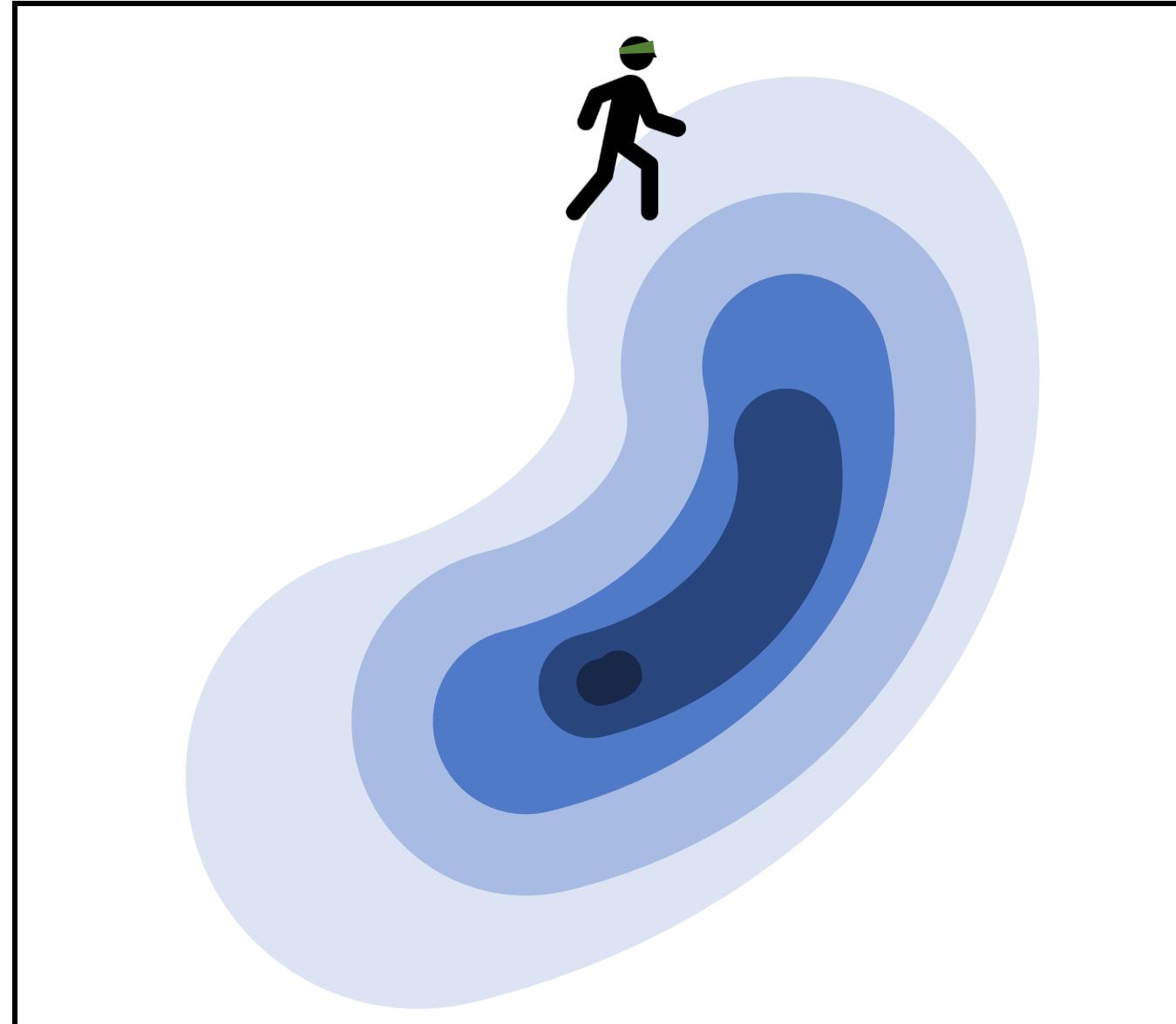
East - West



# Minimizing Objective Function

North - South

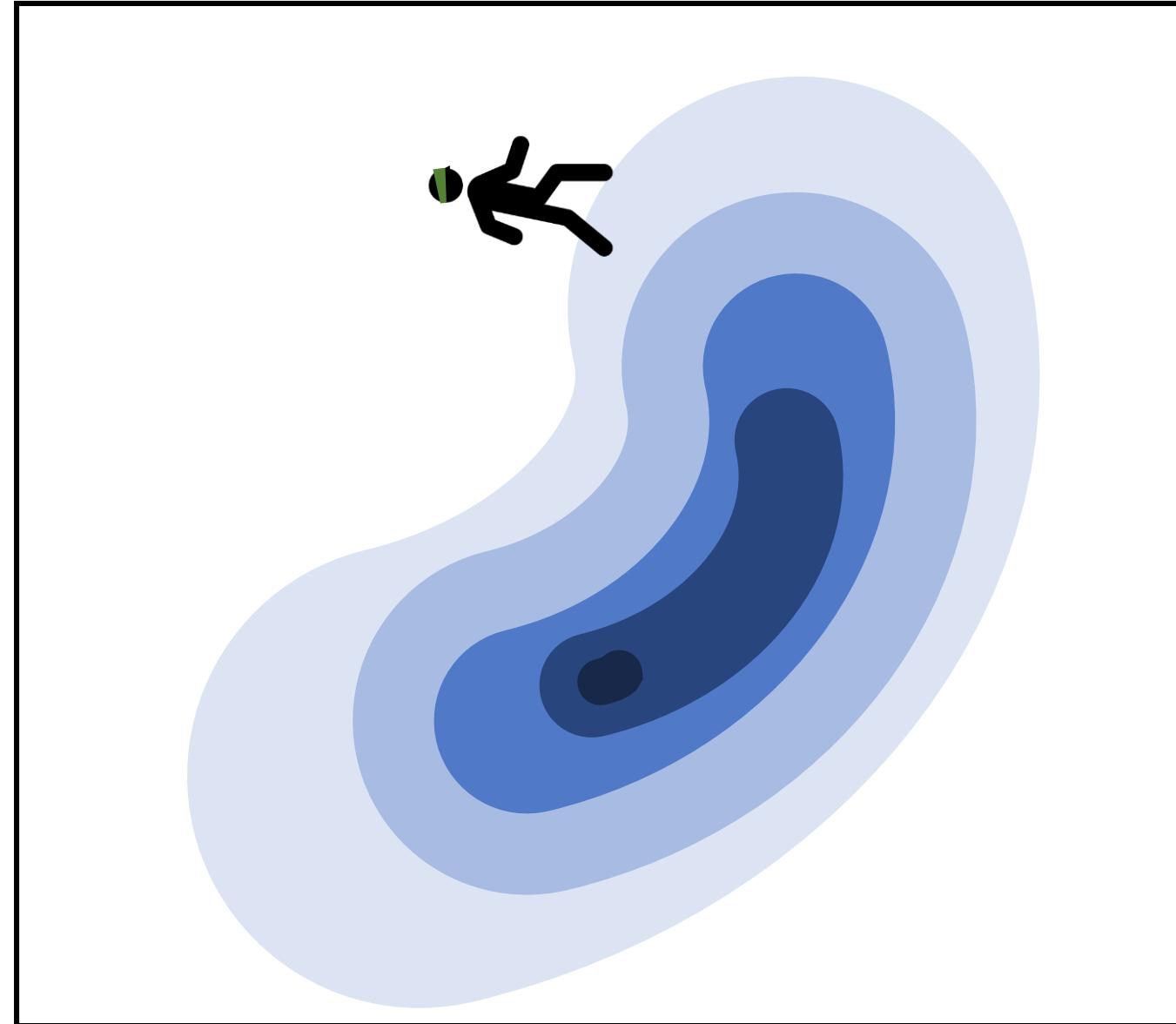
East - West



# Minimizing Objective Function

North - South

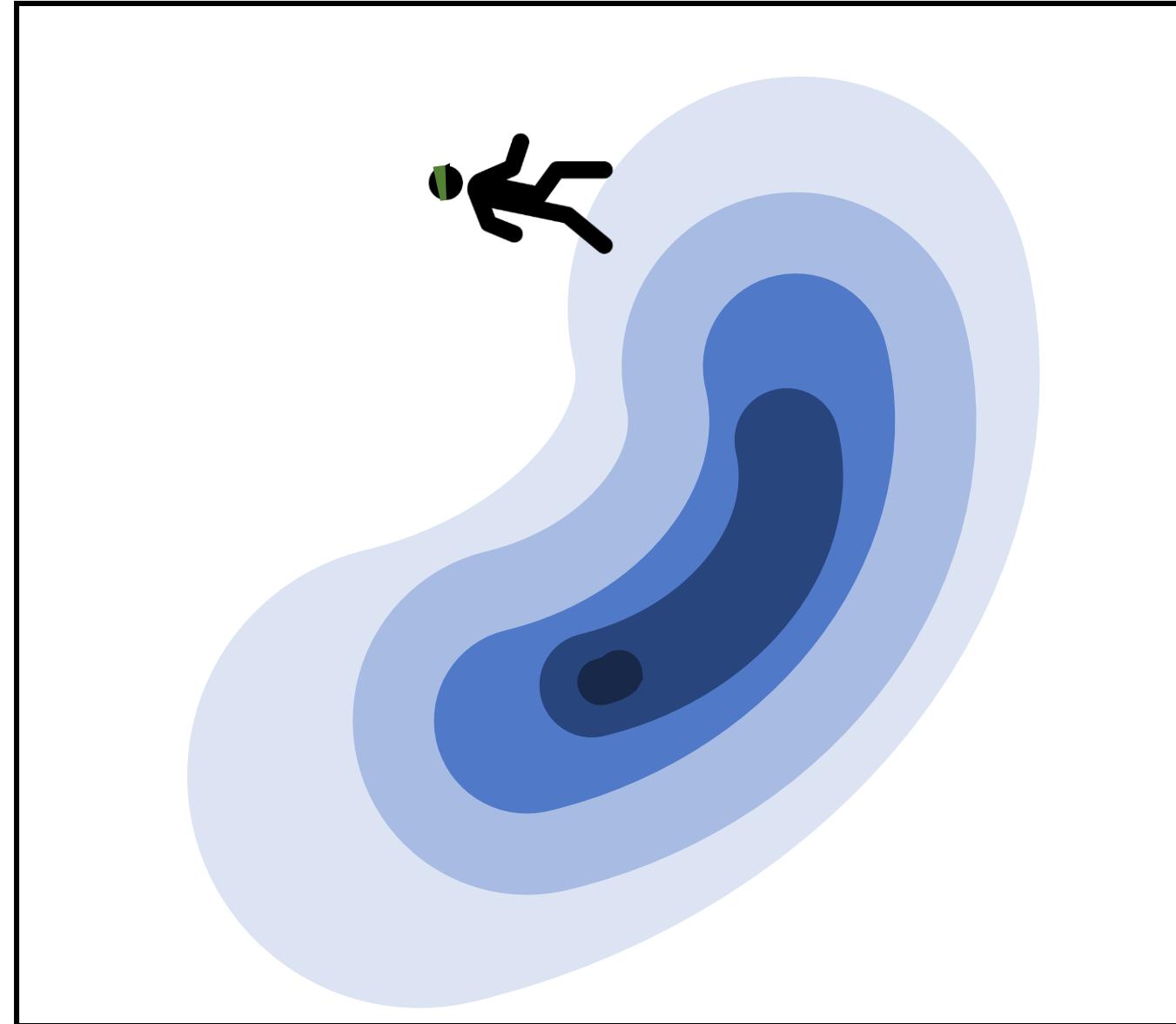
East - West



# Minimizing Objective Function

North - South

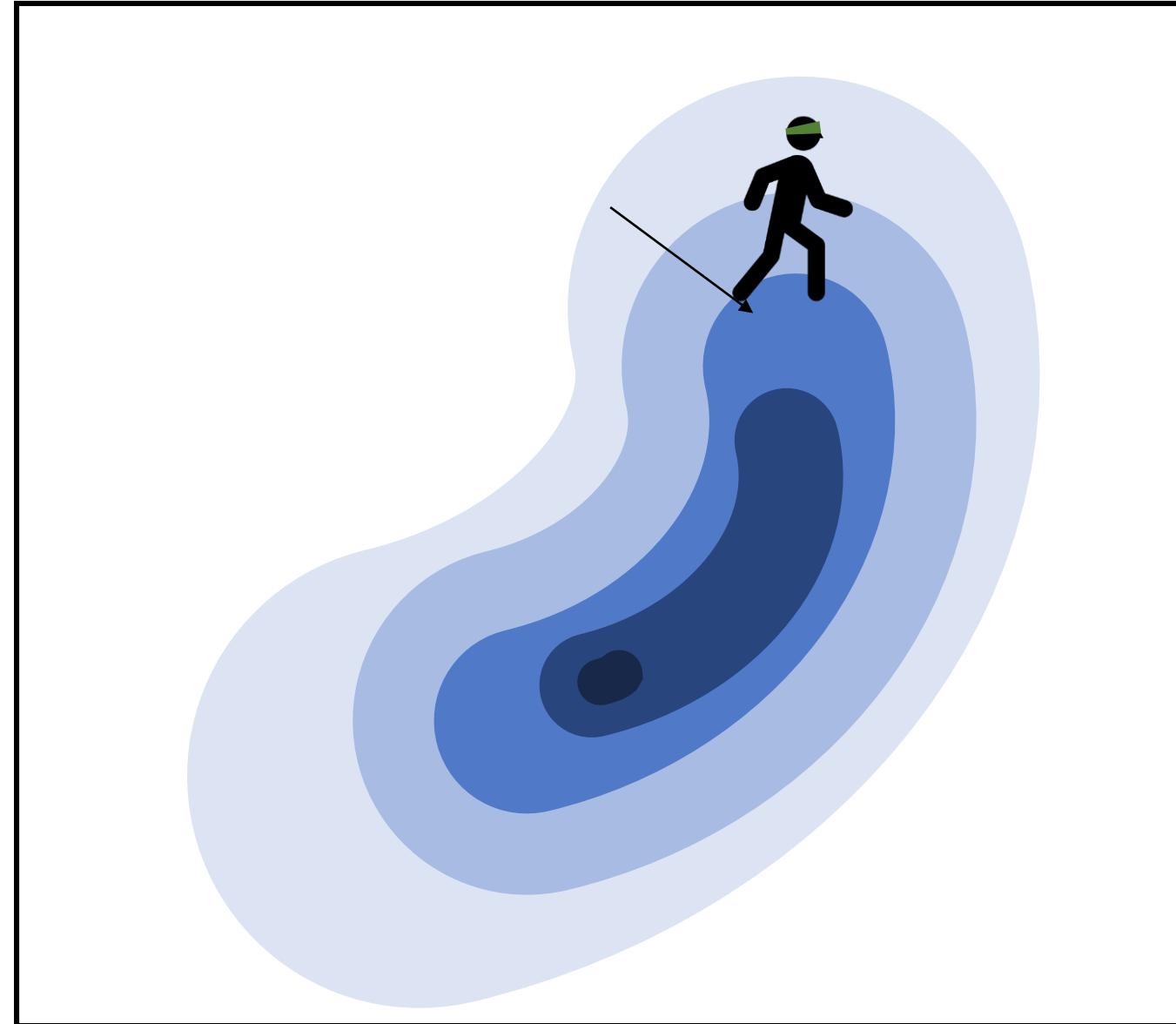
East - West



# Minimizing Objective Function

North - South

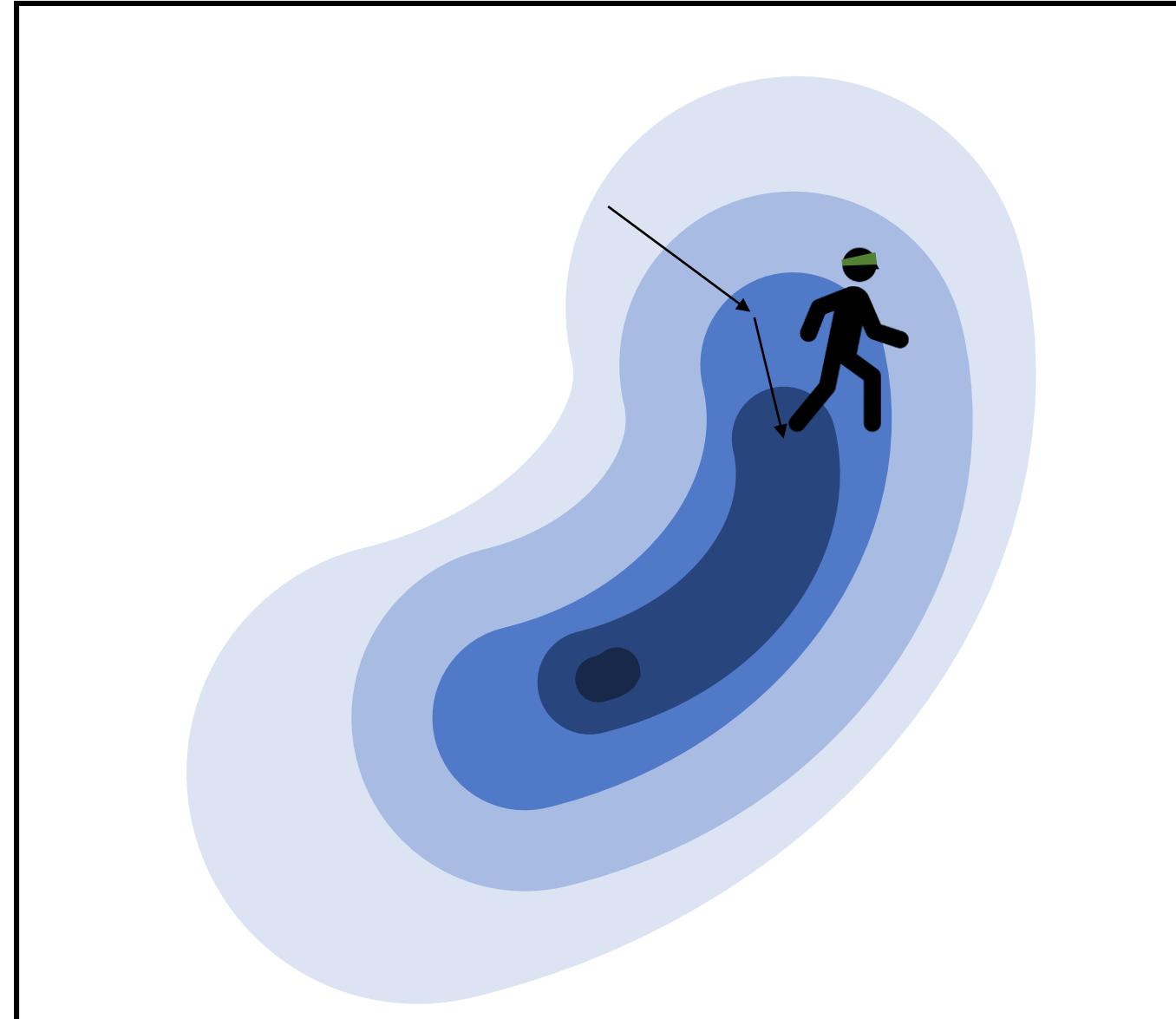
East - West



# Minimizing Objective Function

North - South

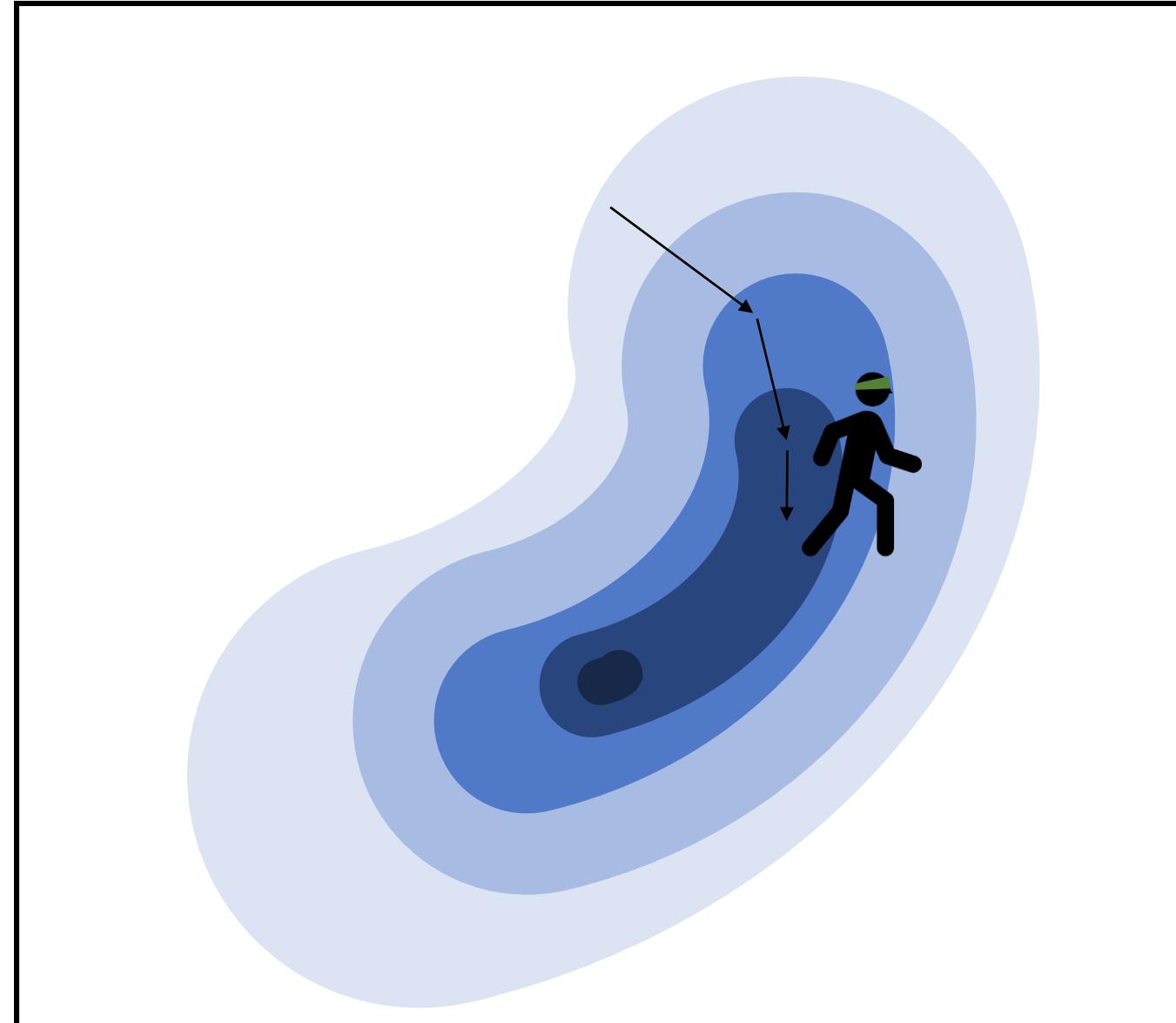
East - West



# Minimizing Objective Function

North - South

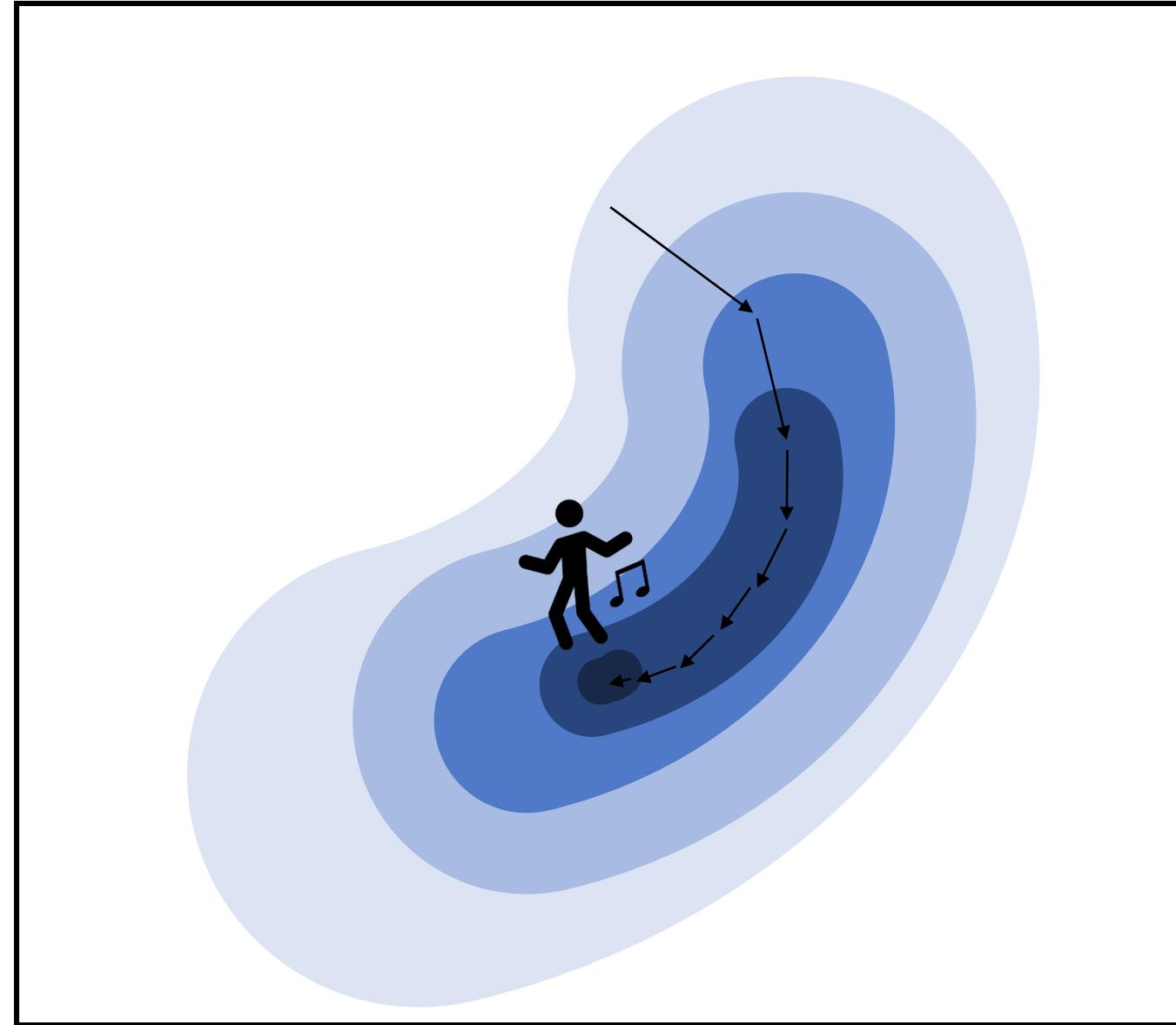
East - West



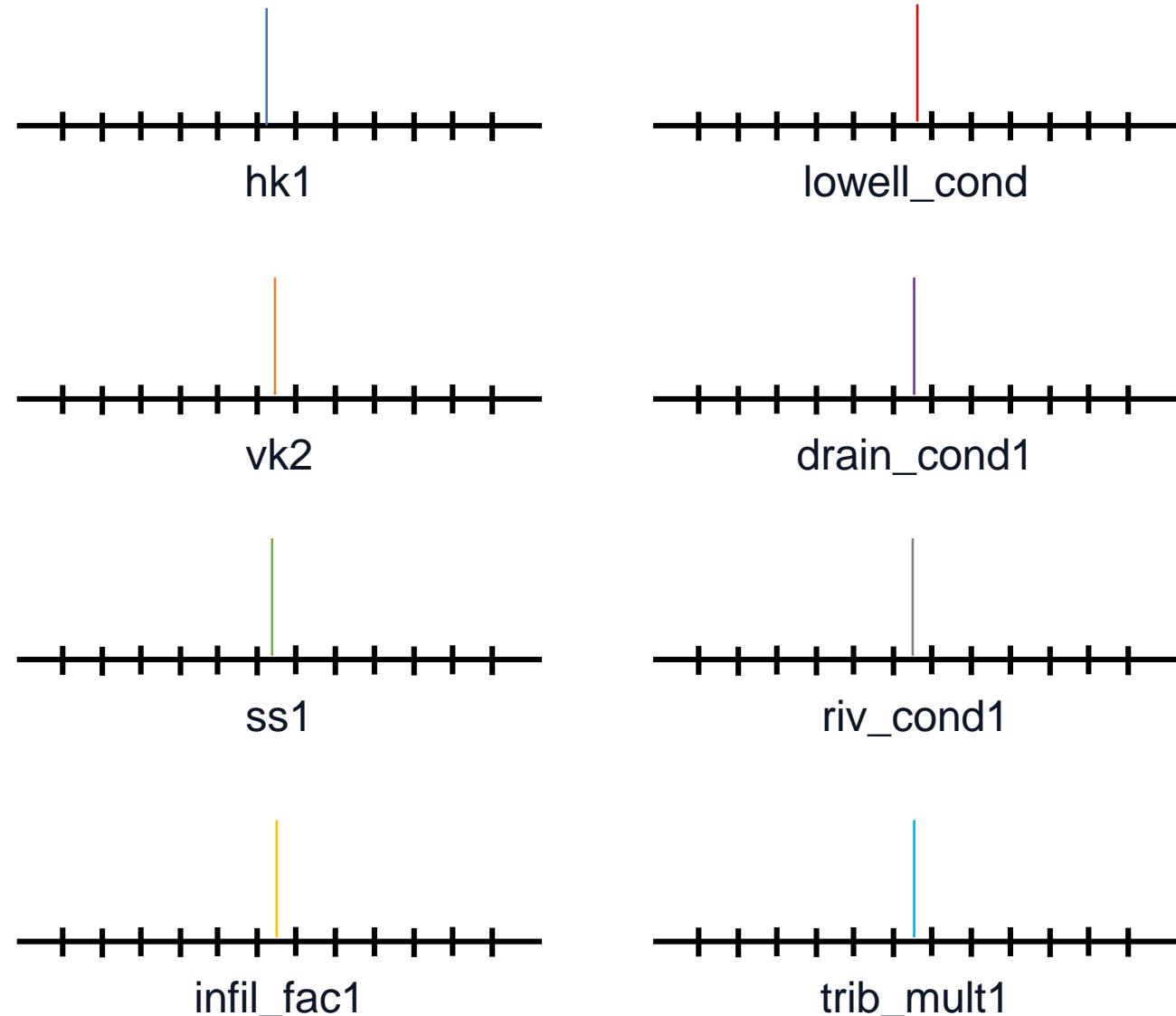
# Minimizing Objective Function

North - South

East - West

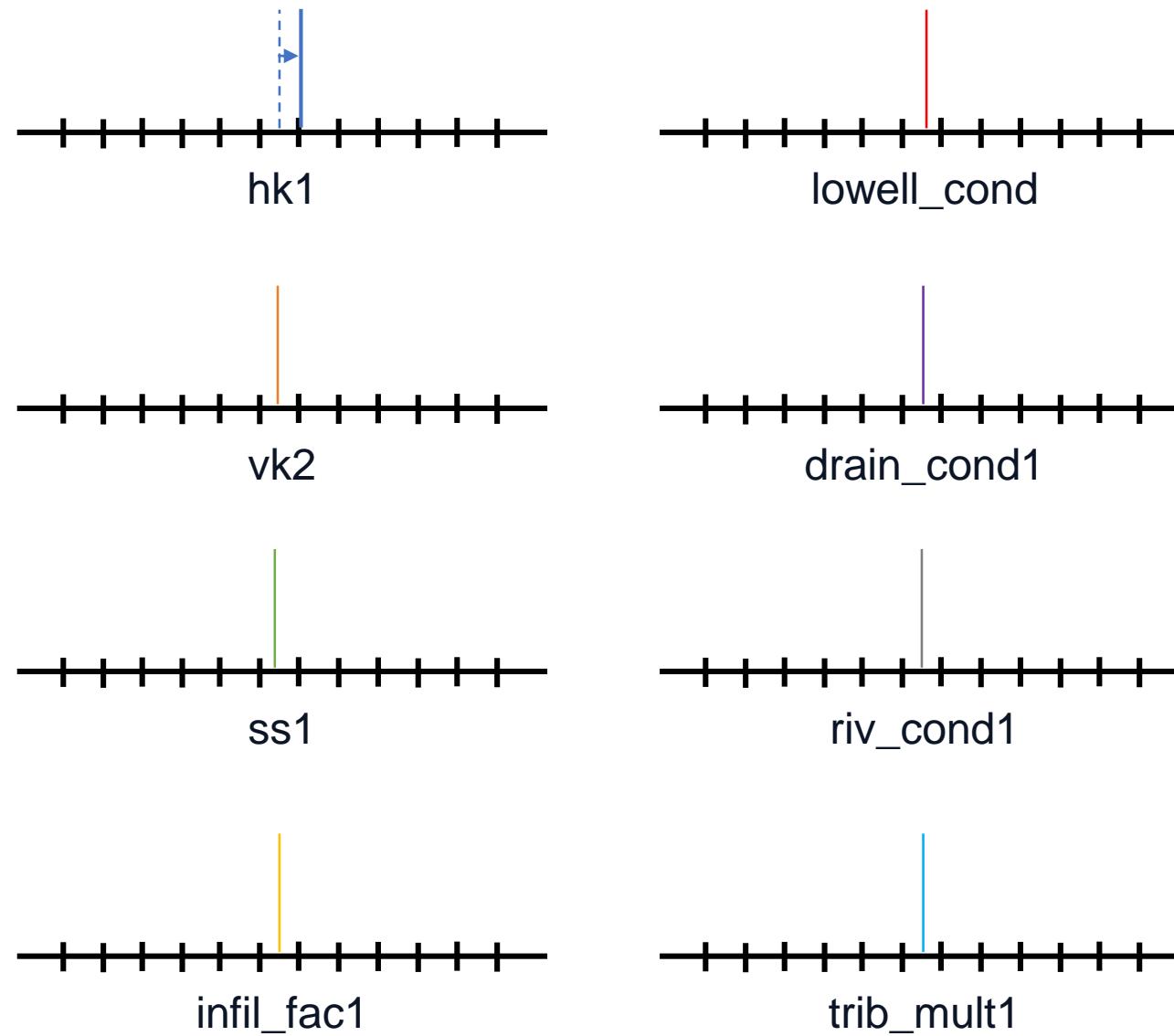


# Minimizing Objective Function



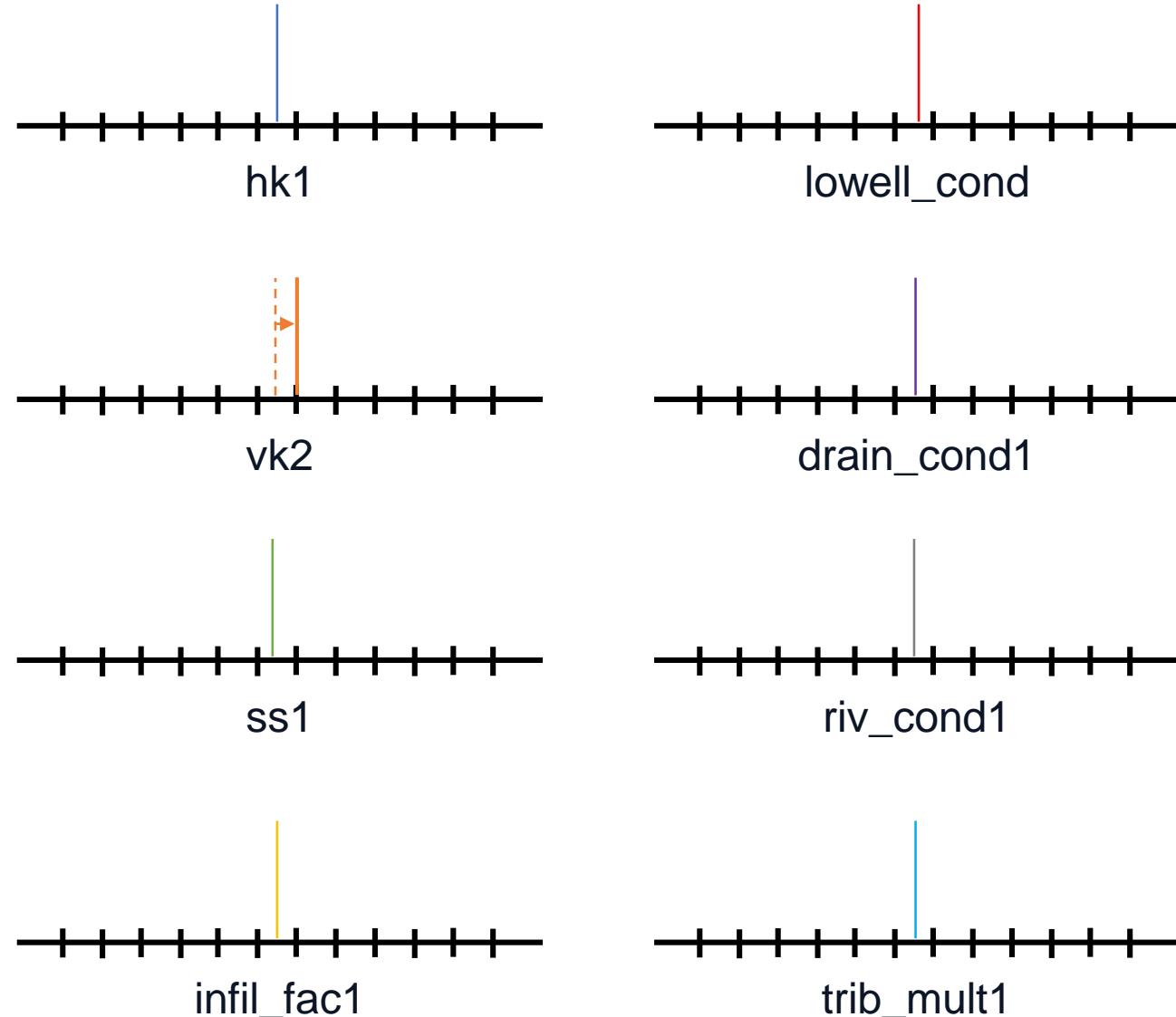
$\Phi$

# Minimizing Objective Function



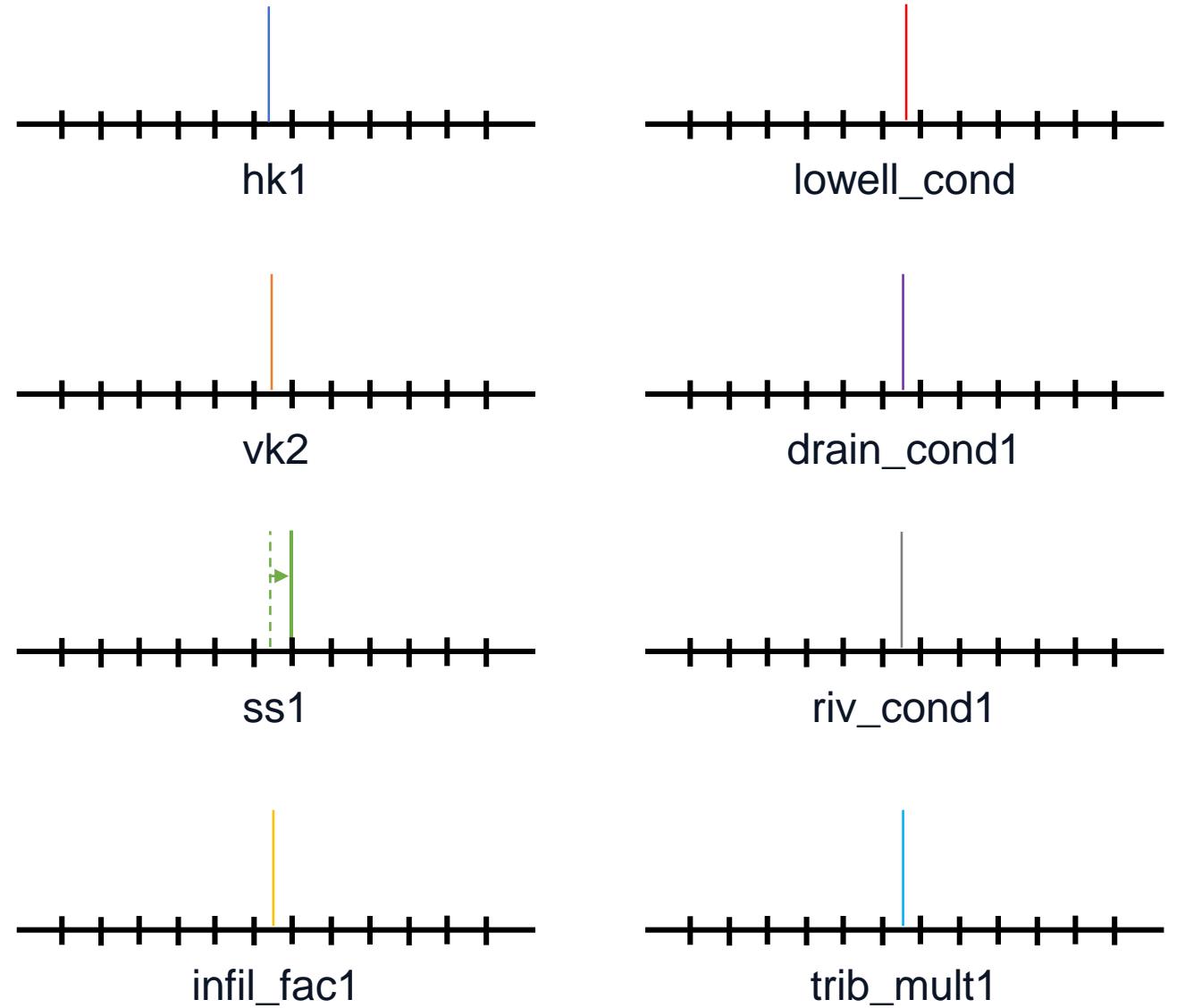
$\Phi \uparrow$

# Minimizing Objective Function



$\Phi \uparrow$

# Minimizing Objective Function

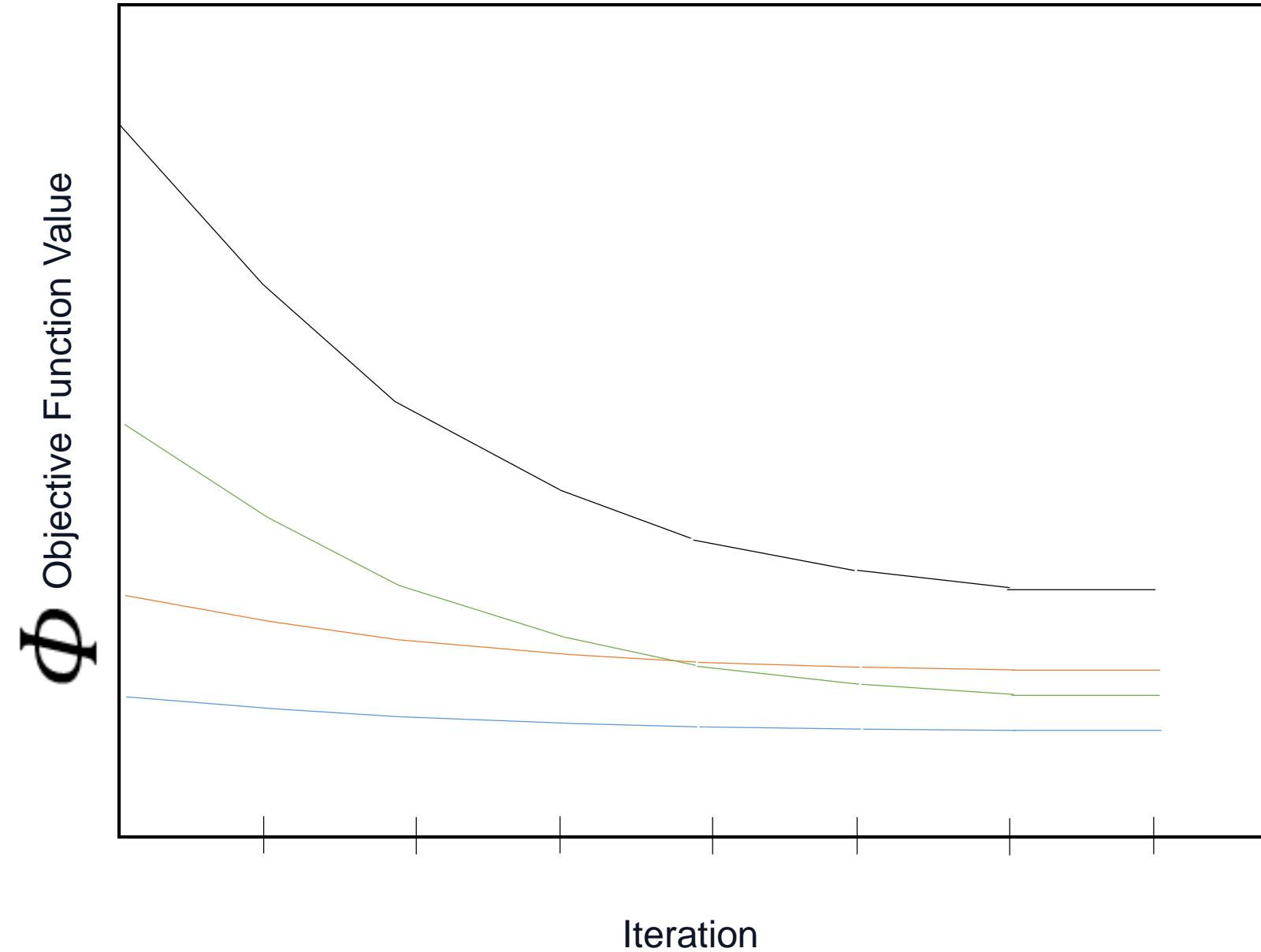


$\Phi \downarrow$

# Minimizing Objective Function

	obs1	obs2	obs3	...	
par1	$\frac{\Delta\phi_1}{\Delta par_1}$	$\frac{\Delta\phi_2}{\Delta par_1}$	$\frac{\Delta\phi_3}{\Delta par_1}$		$\frac{\Delta\phi}{\Delta par_1}$
par2	$\frac{\Delta\phi_1}{\Delta par_2}$	$\frac{\Delta\phi_2}{\Delta par_2}$	$\frac{\Delta\phi_3}{\Delta par_2}$		$\frac{\Delta\phi}{\Delta par_2}$
par3	$\frac{\Delta\phi_1}{\Delta par_3}$	$\frac{\Delta\phi_2}{\Delta par_3}$	$\frac{\Delta\phi_3}{\Delta par_3}$	⋮	$\frac{\Delta\phi}{\Delta par_3}$
par4	$\frac{\Delta\phi_1}{\Delta par_4}$	$\frac{\Delta\phi_2}{\Delta par_4}$	$\frac{\Delta\phi_3}{\Delta par_4}$	⋮	$\frac{\Delta\phi}{\Delta par_4}$
⋮	⋮	⋮	⋮	⋮	

# Minimizing Objective Function



# Common Challenges

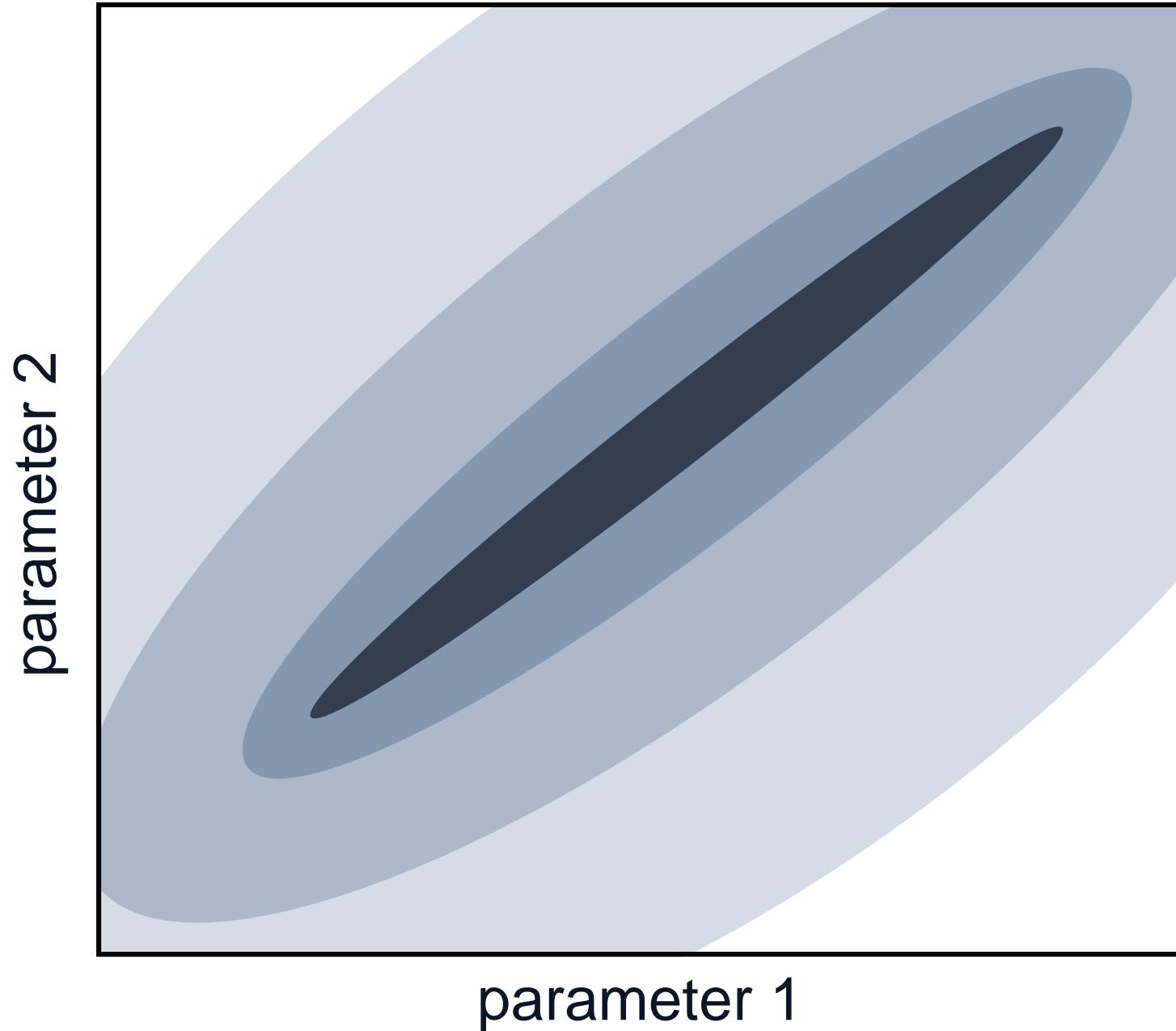
## long run times

COMPUTING JACOBIAN:

Iteration type: base parameter solution  
calculating jacobian... running model 1488 times

# Common Challenges

parameter correlation



# Common Challenges

non-uniqueness

$$Q = K \frac{\Delta H}{L}$$

recharge

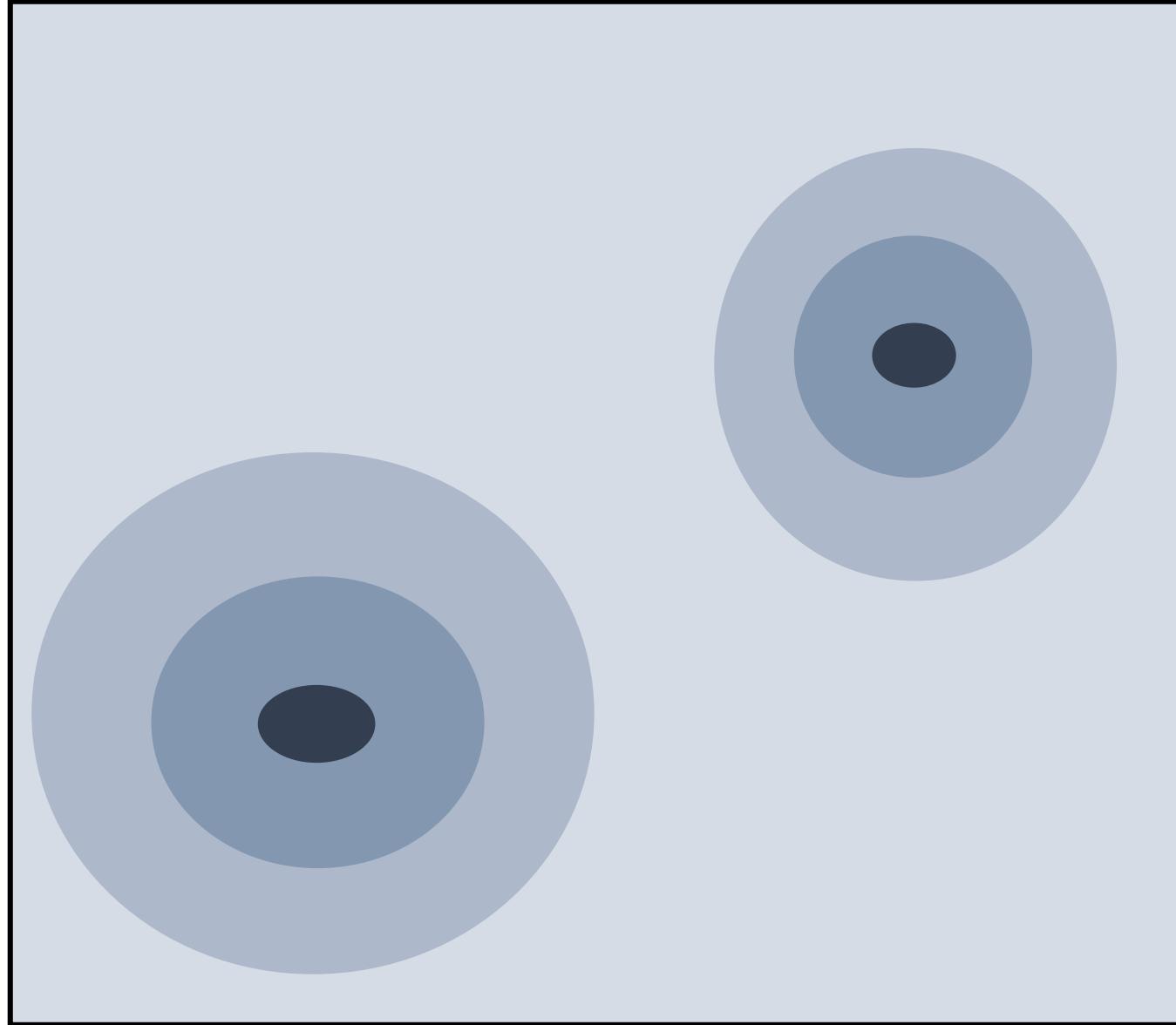
hydraulic conductivity

# Common Challenges

non-uniqueness

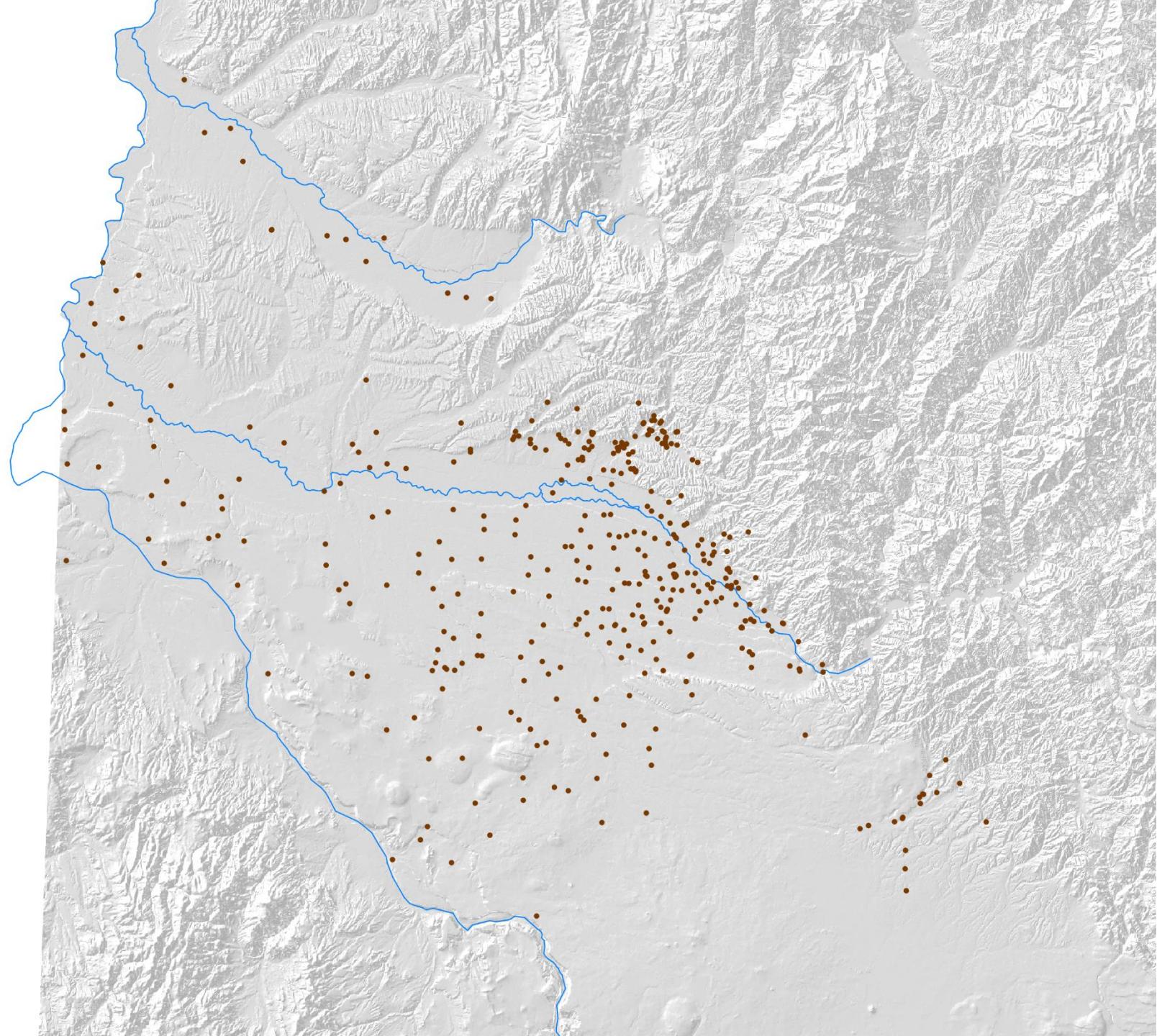
parameter 2

parameter 1



# Common Challenges

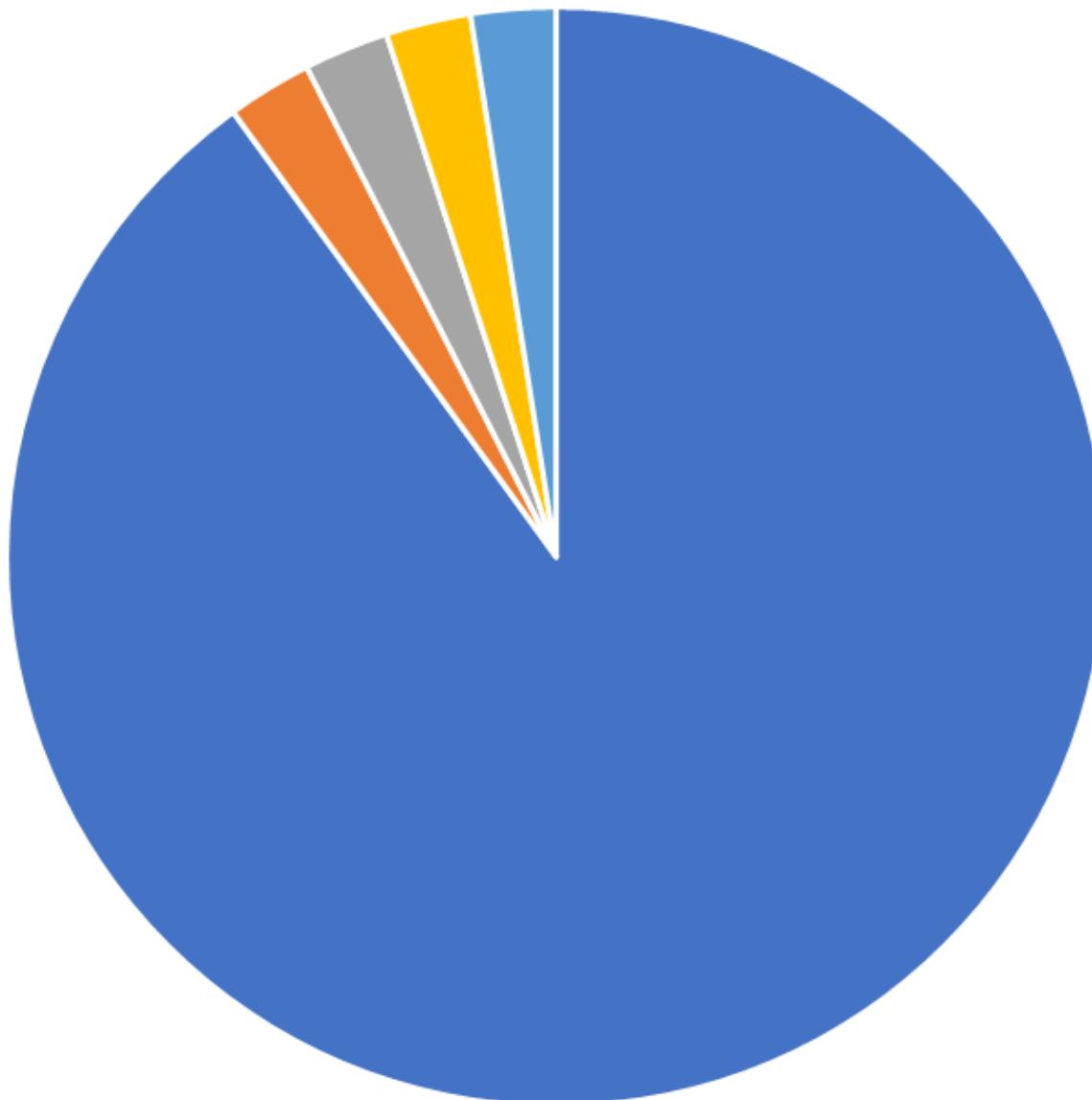
weighting



# Common Challenges

weighting

objective function

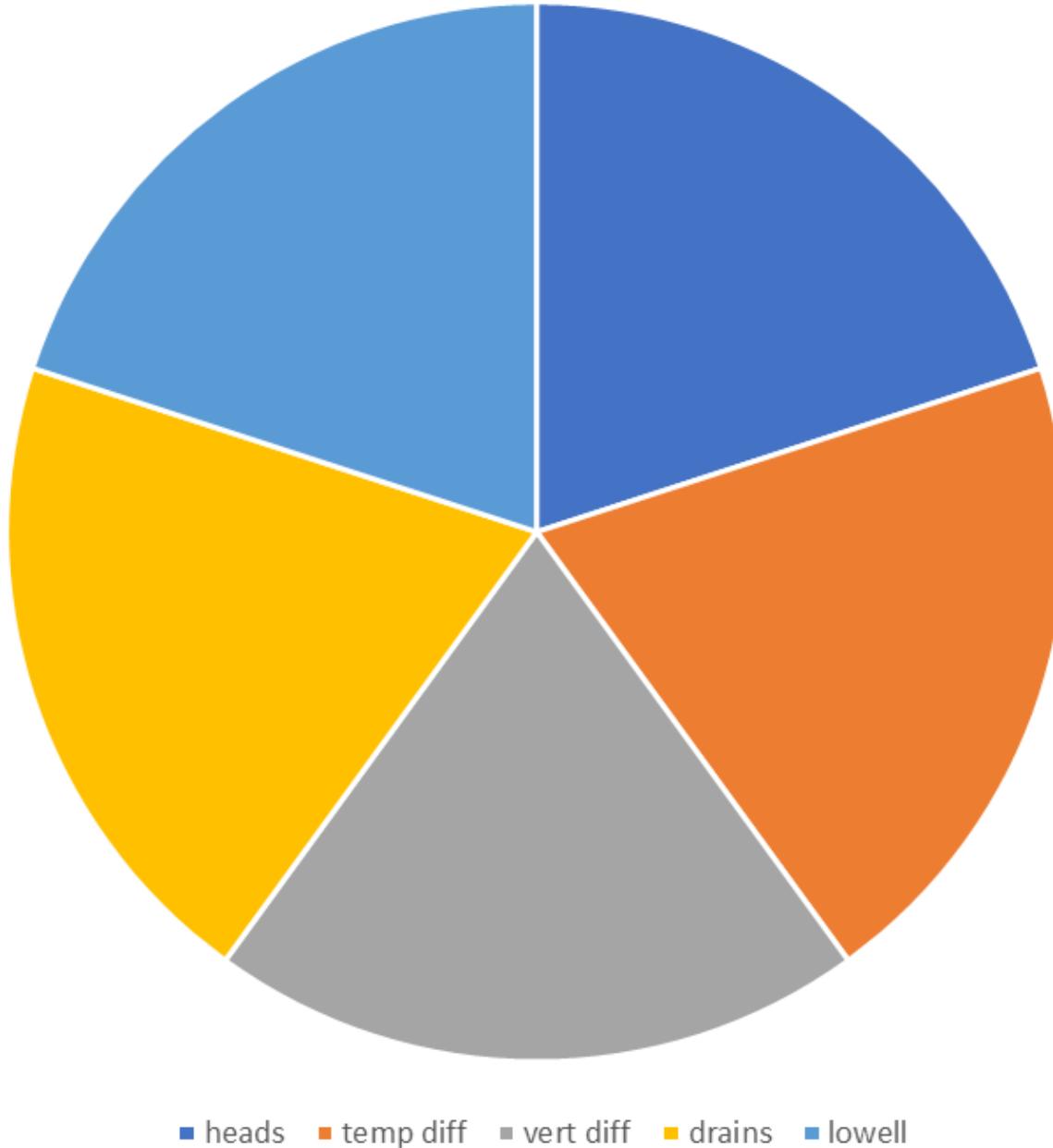


■ heads ■ temp diff ■ vert diff ■ drains ■ lowell

# Common Challenges

weighting

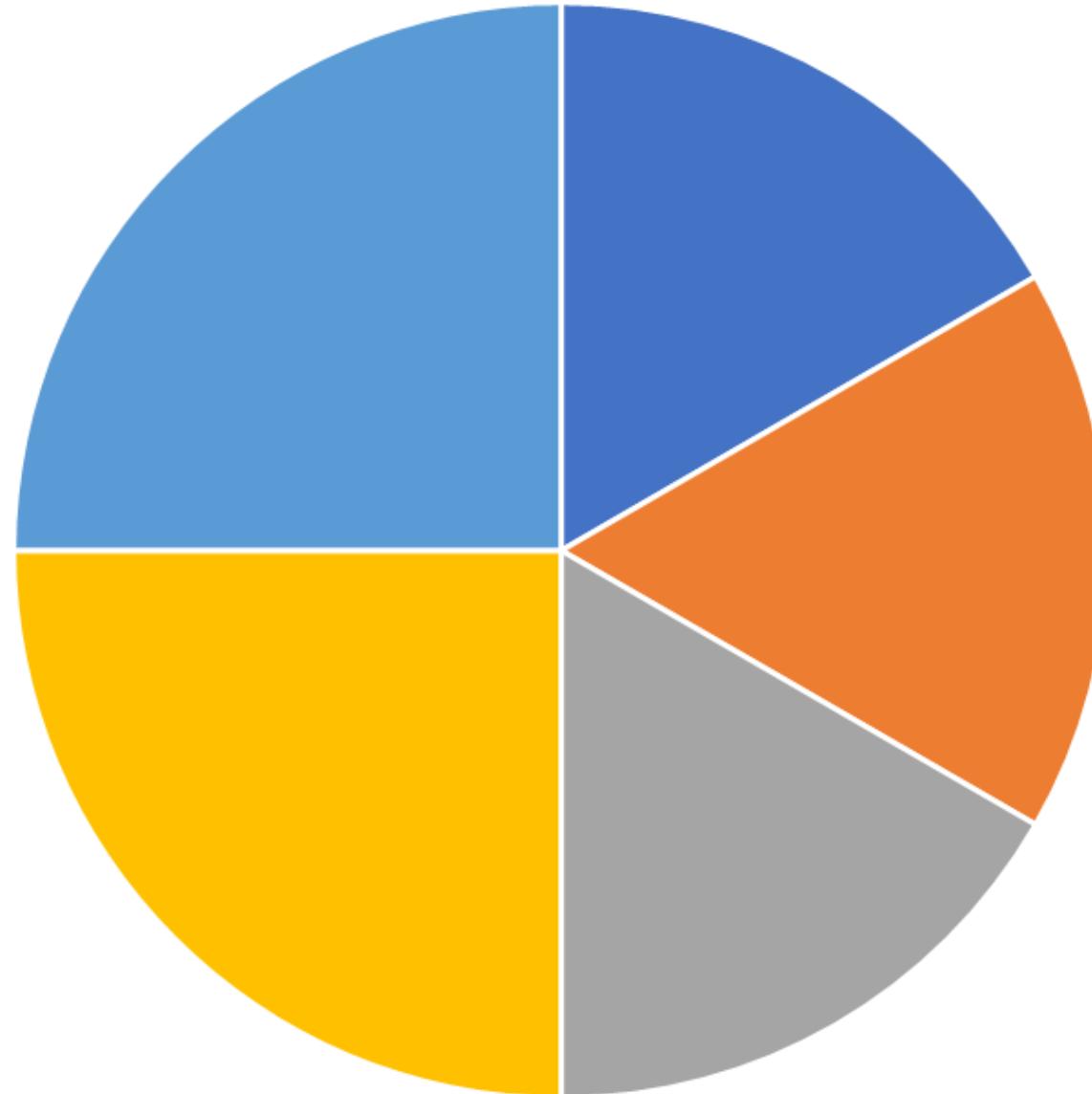
objective function



# Common Challenges

weighting

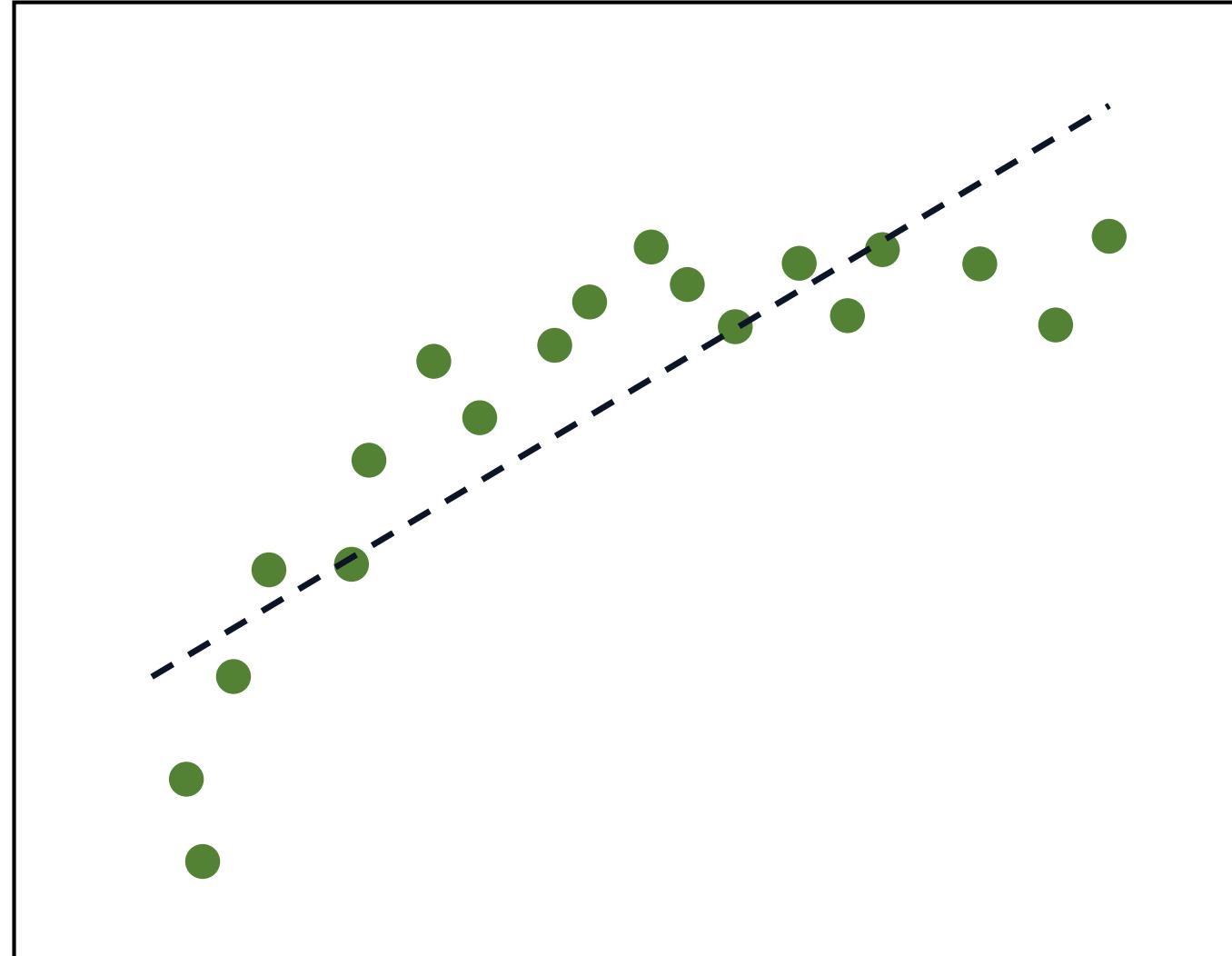
objective function



■ heads ■ temp diff ■ vert diff ■ drains ■ lowell

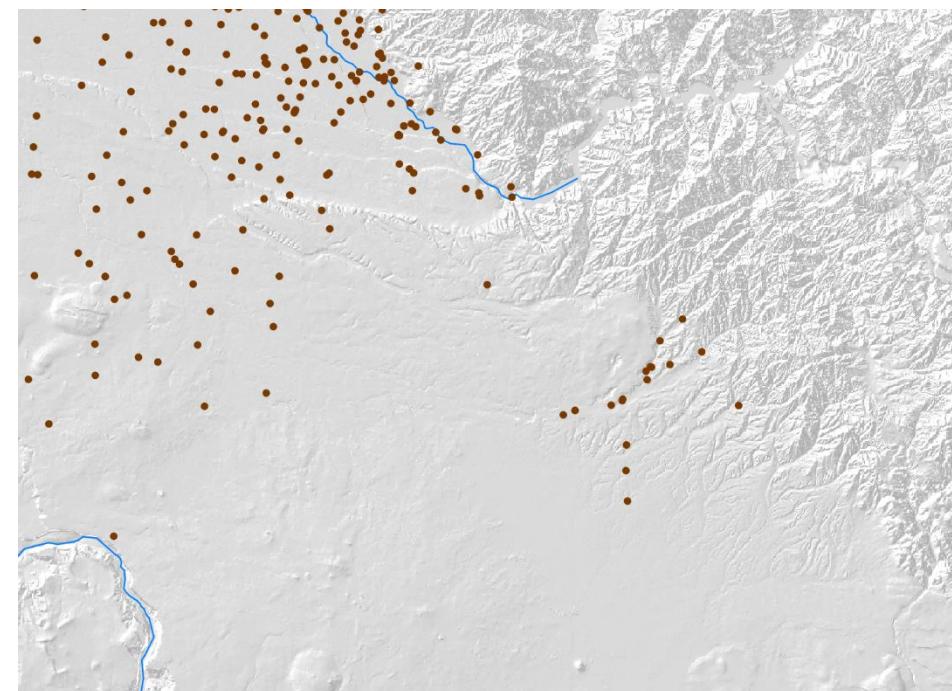
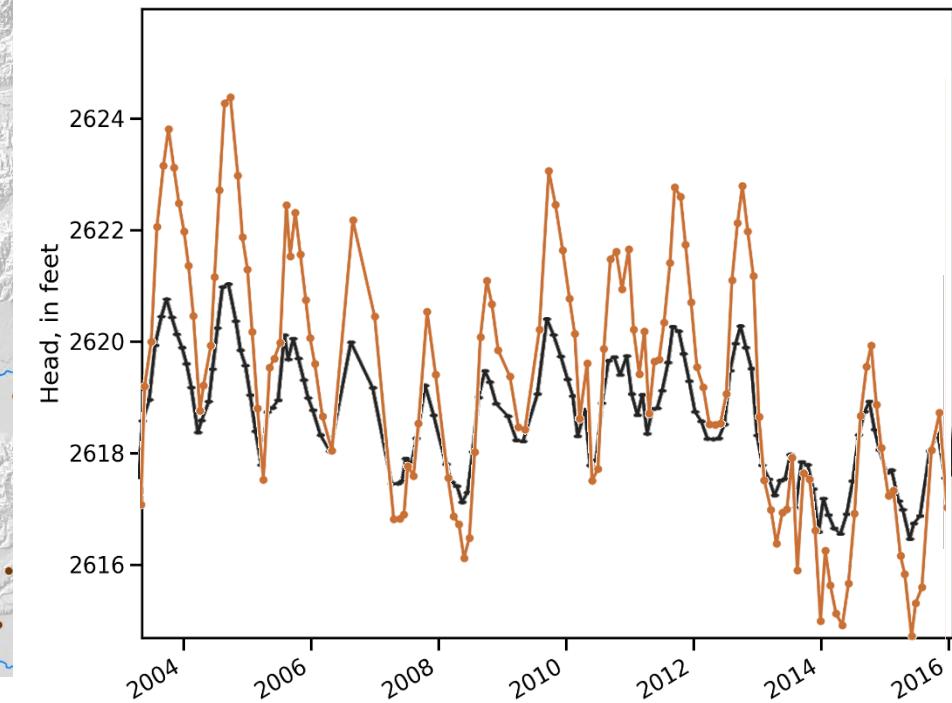
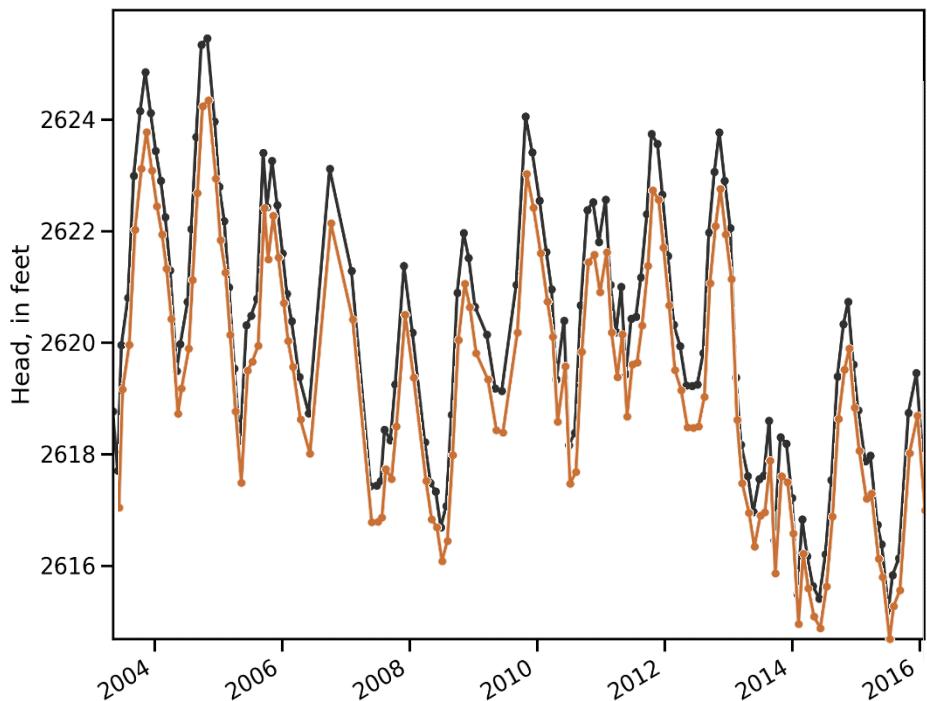
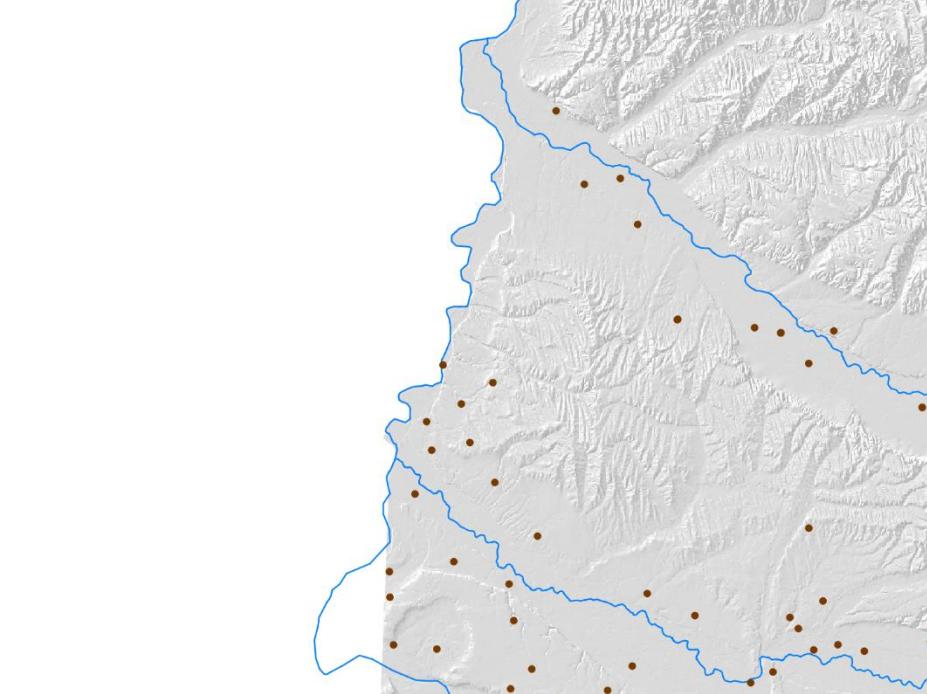
# Common Challenges

model  
inadequacy



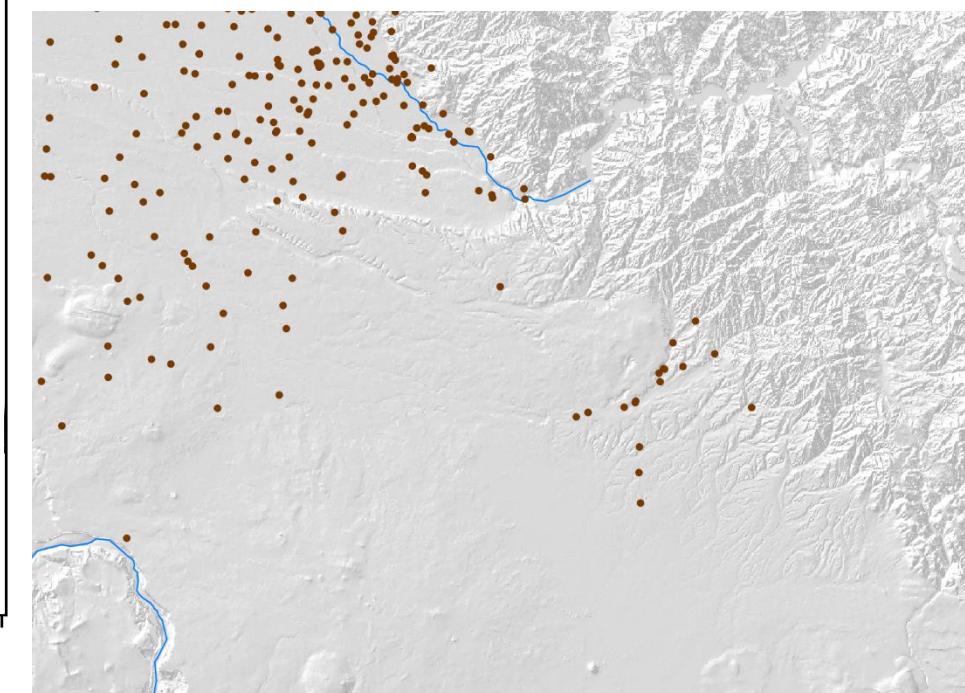
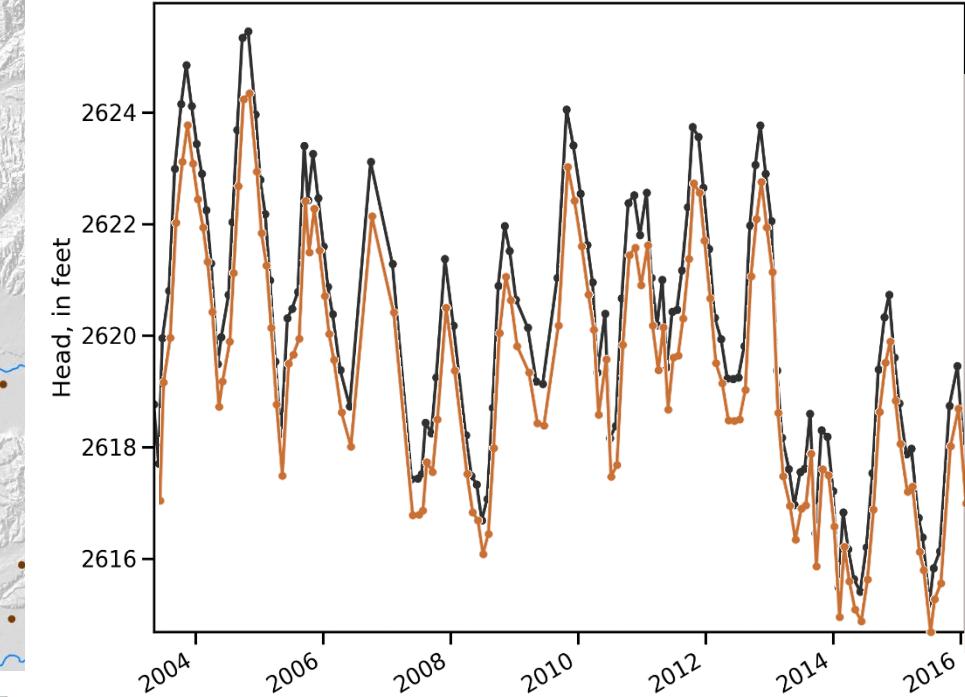
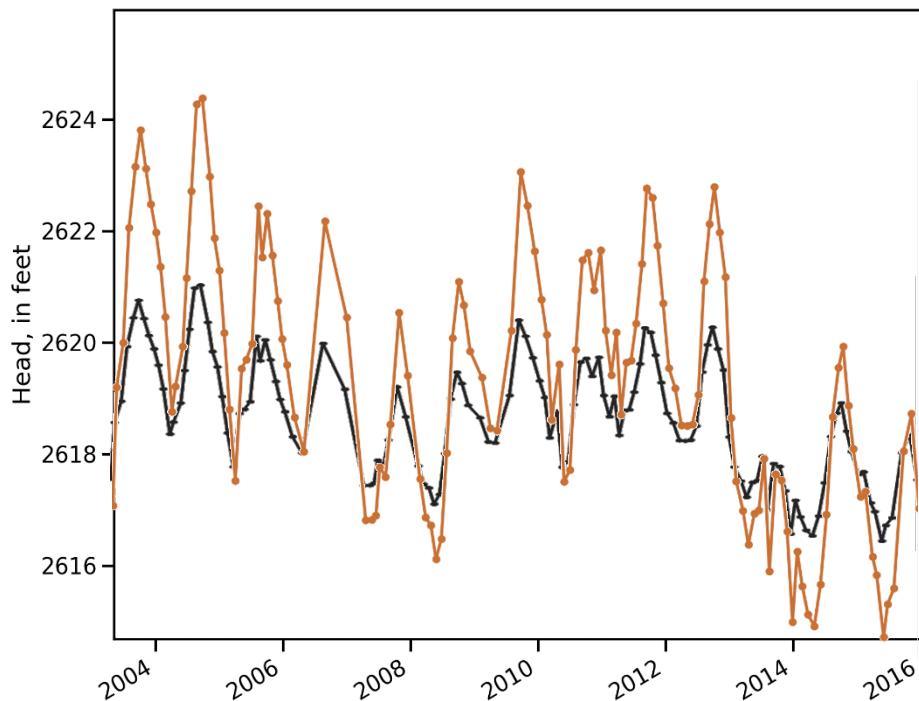
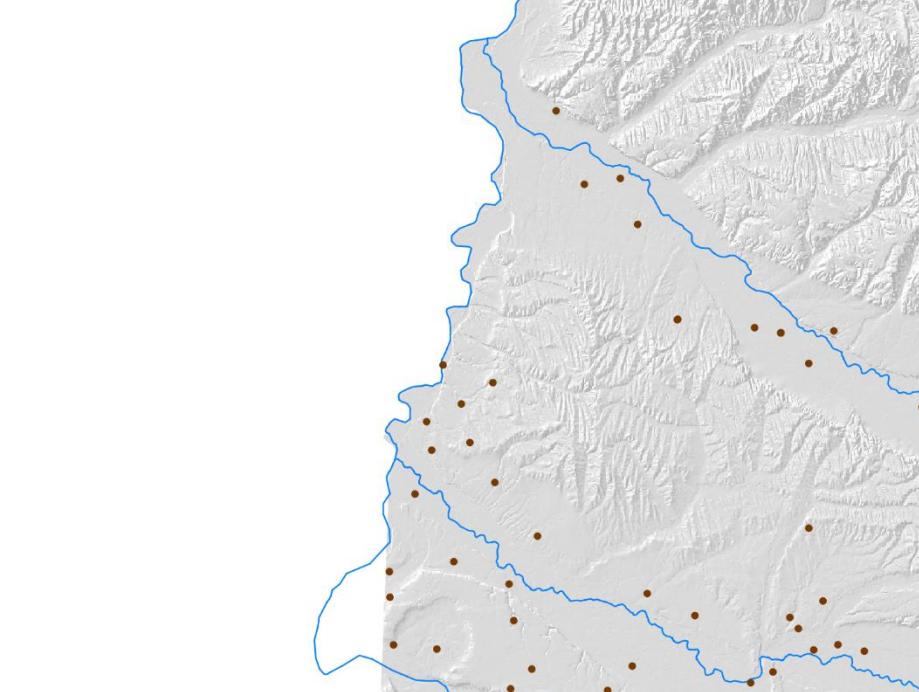
# Common Challenges

systematic  
misfits or  
tradeoffs



# Common Challenges

## systematic misfits or tradeoffs



# Common Challenges

overfitting

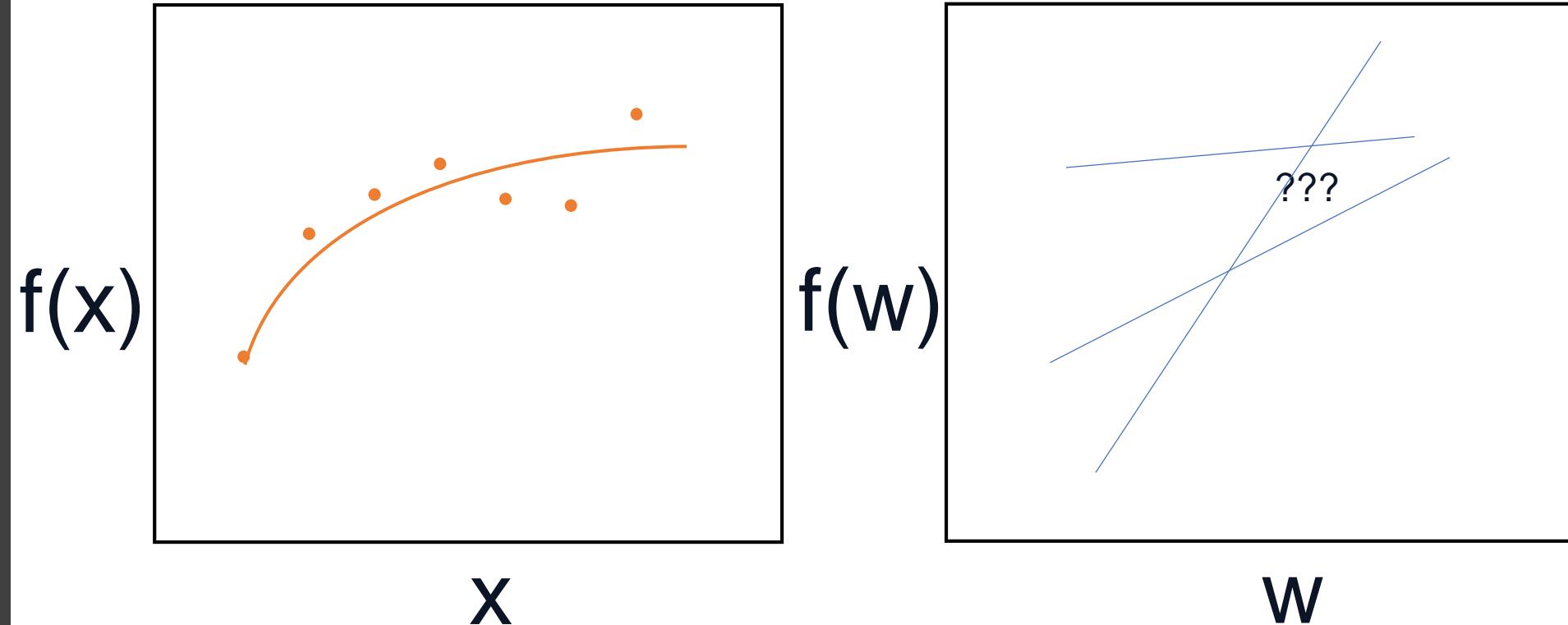
$f(x)$

$x$



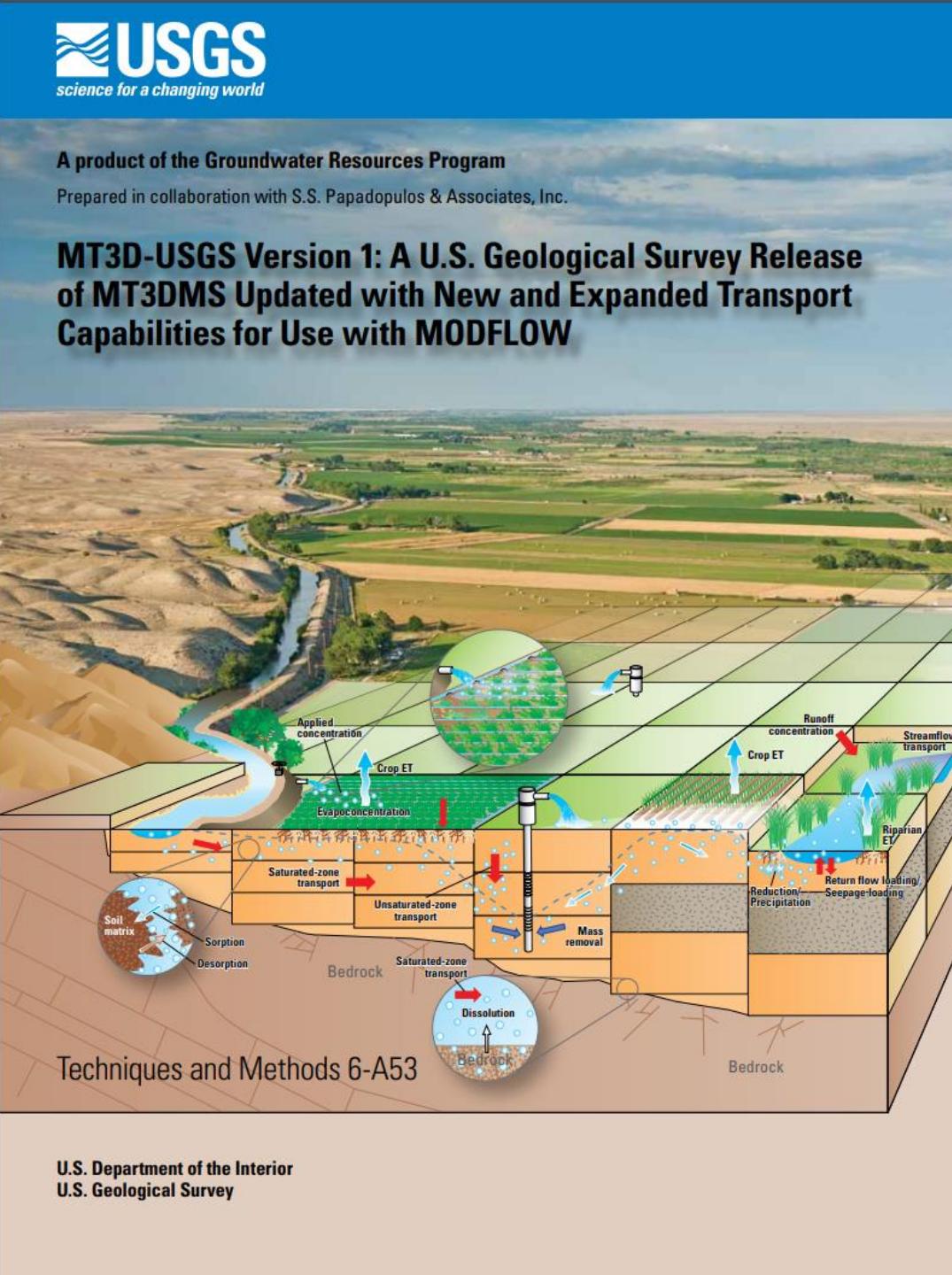
# Common Challenges

scenario forecasts uninformed by data



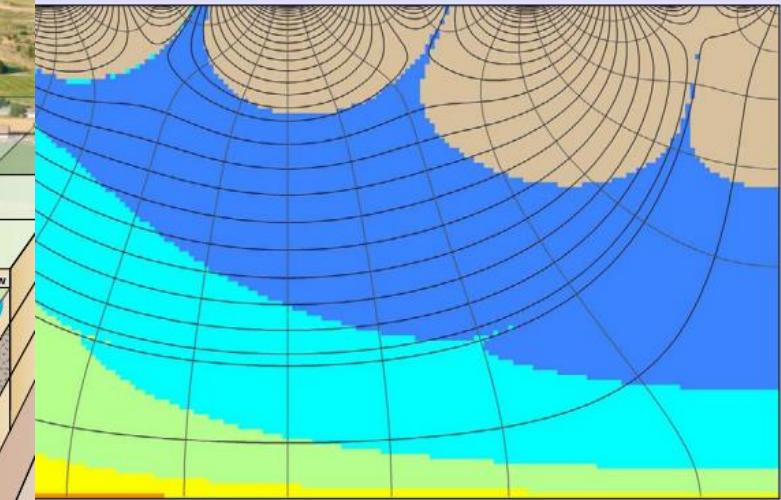
# Common Challenges

scenario forecasts uninformed by data



or MODPATH Version 6—  
Tracing Model for MODFLOW

Techniques



Methods 6-A41

rior

# Implementation

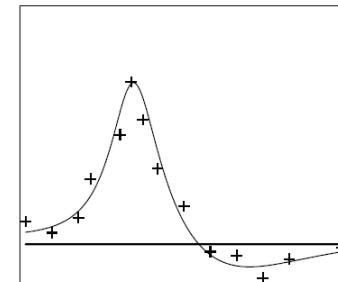
# PEST

## Model-Independent Parameter Estimation

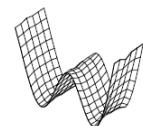
### User Manual Part I:

### PEST, SENSAN and Global Optimisers

(See part II for documentation of PEST support and uncertainty analysis utilities.)



7th Edition published in 2018  
Latest additions: May 2019

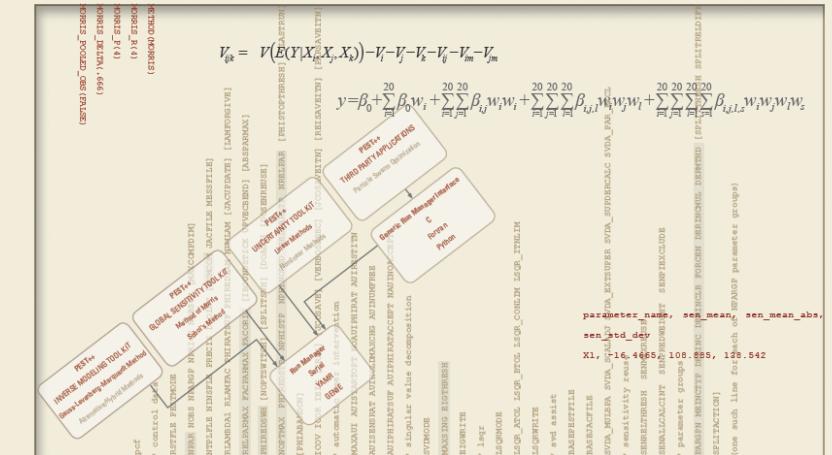


Watermark Numerical Computing



Groundwater Resources Program  
Prepared in cooperation with U.S. Environmental Protection Agency,  
Great Lakes Restoration Initiative

## Approaches in Highly Parameterized Inversion: PEST++ Version 3, A Parameter ESTimation and Uncertainty Analysis Software Suite Optimized for Large Environmental Models



Techniques and Methods 7-C12

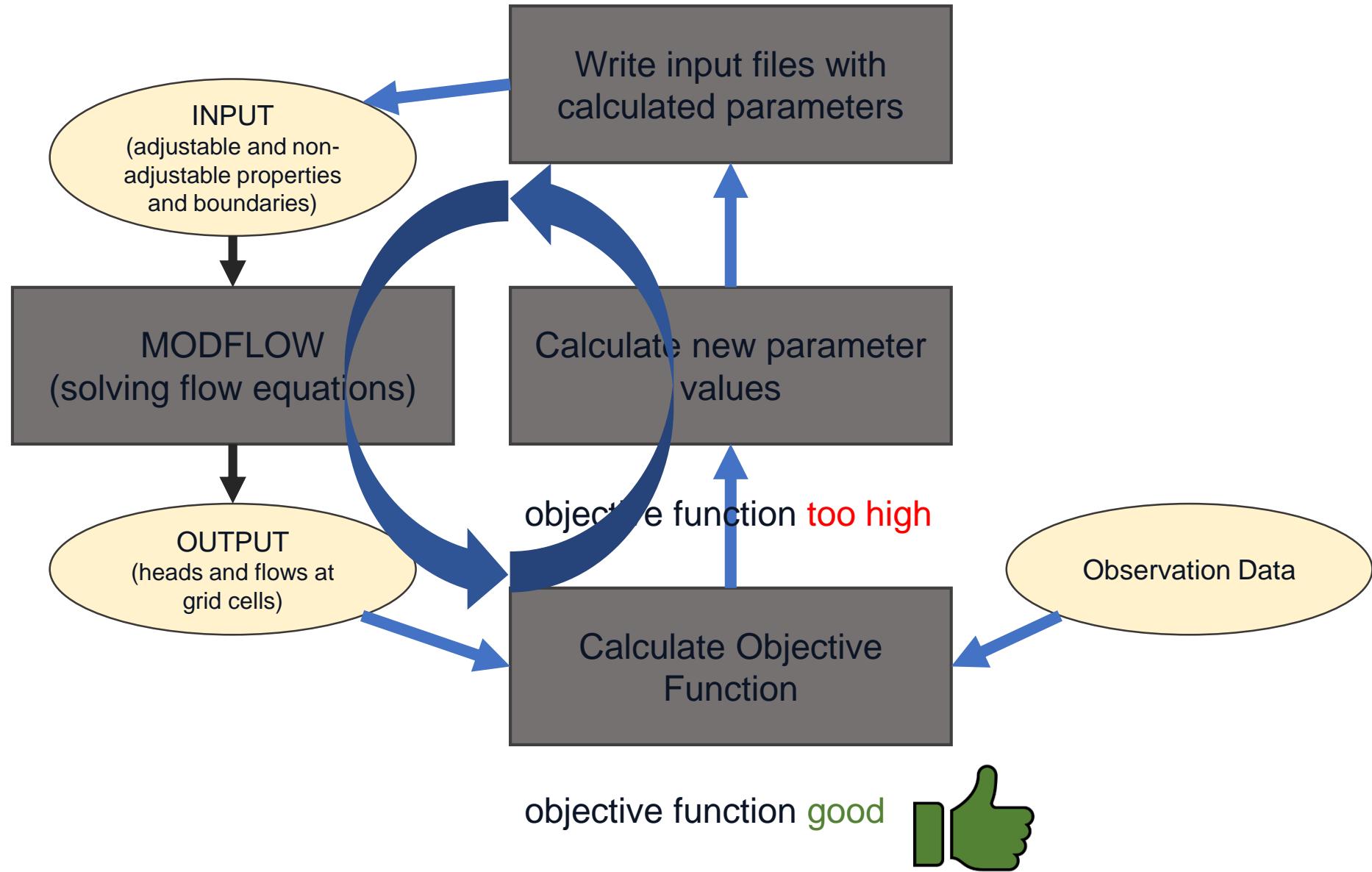
U.S. Department of the Interior  
U.S. Geological Survey

# Wrapping Model

MODEL

PEST

DATA



# Techniques

many settings for  
tweaking parameter  
estimation

Tikhonov Regularization

Singular Value Decomposition

SVD-Assist

Iterative Ensemble Smoother

# High-throughput computing

Generalized Run Manager Interface  
by:

The PEST++ Development Team

starting PANTHER master...

IP addresses:

0.0.0.0:4004 (IPv4)

PANTHER master listening on socket: 0.0.0.0:4004 (IPv4)

COMPUTING JACOBIAN:

Iteration type: base parameter solution  
calculating jacobian... running model 1488 times

waiting for agents to appear...

PANTHER progress

runs(C = completed | F = failed | T = timed out)  
agents(R = running | W = waiting | U = unavailable)

# High-throughput computing

Machine Access

## HPC SYSTEMS

### General Purpose HPC

- Good place to start
- CPU and GPU
- 143 nodes
- 3,728 CPU, 56,596 CUDA
- 100 Tflop/s

Yeti



### Flagship System

- Large-scale Models
- CPU Only
- 232 nodes
- 9,280 CPU (18,560 hyper-threads)
- 448 Tflop/s

Denali



### Prototype System

- ML and Analytics at scale
- Built-in Software Stack
- 22 nodes, 792 CPU
- 122,800 CUDA
- 15,360 Tensor

Tallgrass



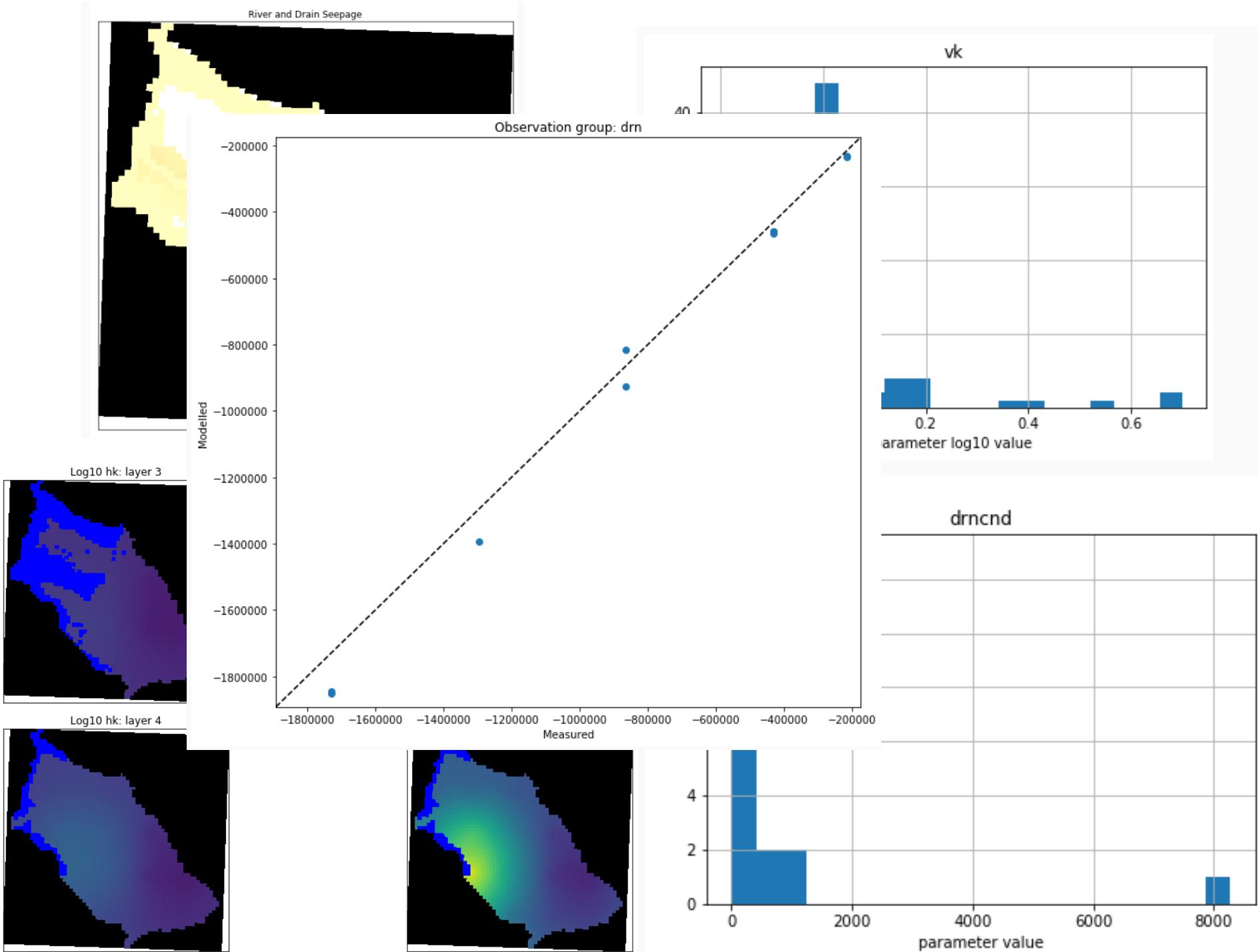
### Tiered Storage System

- 4.2 PB Total
- Connects to Denali & Tallgrass
- Provide tiers of storage (high performance & object)

Caldera



# Iterate

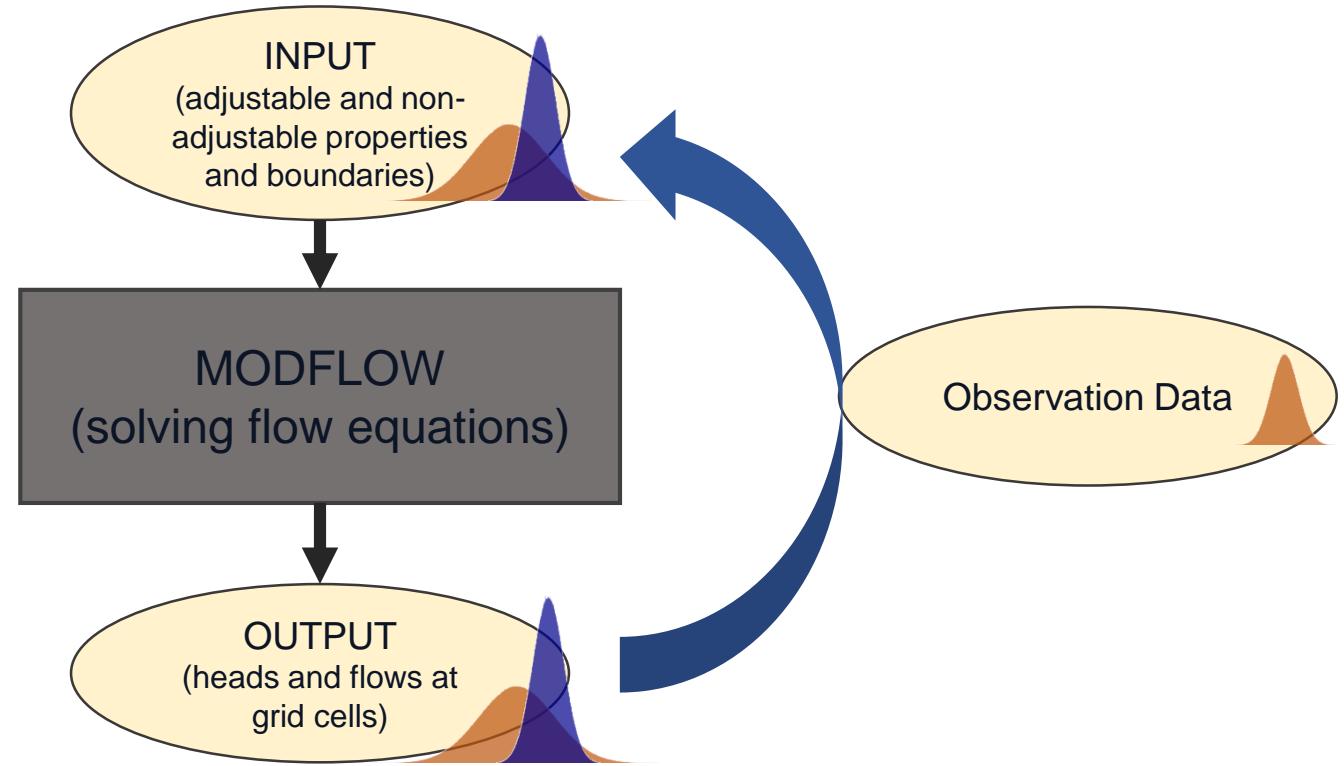


# Iterate

```
# settings
# model
sim_workspace = 'base'
model_name = 'mf6-tv_hist_ss'
ss_years = list(range(1998, 2007))
# pest
read_parameters_from_file = False
read_parameters_group_mean = False
balance_group_phi = True
tie_groups = ['leakdistprop', 'leakfac']
# tie_groups = ['drncnd', 'hk', 'infilfac', 'Leakdistprop', 'Leak
man_obs_group_weight_mult = pd.DataFrame({
    'obgnme' : ['obs_well', 'drn', 'lowell', 'vert_head_diff'],
    # 'weight_mult' : [1, 1e-2, 5e-4, 1]
    'weight_mult' : [1, 3e2, 1e-1, 1e1]
})
```

# What else?

# Uncertainty Analysis



Thanks for listening!